

GSVA for mutil Group

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```
library(tibble)
library(tidyr)
library(dplyr)
library(rtracklayer)
```

```
# load function from local files
source(here::here("source", "DEG_functions.R"))
```

1. Read the count data

In this section, we will read the clean count data from the `synaptosomes_bulkRNA` folder. We will read the data and merge them into a single table. The final table will be stored in `../dataresults/bulkRNA_counts_clean.csv`.

```
input_count <- read.csv(here::here("data", "bulkRNA",
                                   "bulkRNA_counts_cleaned.csv"))

counts <- as.data.frame(input_count) %>%
  column_to_rownames(var = "gene")
colnames(counts) <- gsub("_", "-", colnames(counts))

# raw sample list
sample_list_raw <- read.csv(here::here("data", "bulkRNA",
                                       "sample_info_AD.csv")) %>%
  mutate(condition = paste0(Diagnosis, "_", Treatment),
         sample = gsub("_", "-", Sample.name))

# Ensure the column names of counts exist in Sample.name
new_colnames <- sample_list_raw$Label[match(colnames(counts), sample_list_raw$sample )]

# Assign new column names
colnames(counts) <- new_colnames

# sort the columns by the colname
condition_list <- data.frame(
  group =sample_list_raw$condition
)

row.names(condition_list) <- sample_list_raw$Label

counts<- counts[, rownames(condition_list)]

gene_name_mapping<- readRDS(here::here("data","ref" ,"gene_name_mapping.rds"))
```

2. Differential expression analysis

In this section, we will perform differential expression analysis using DESeq2. We will compare the 22q vs Control in the vehicle condition. The results will be stored in `results/02-DEG-V_10/DESeq2_results.csv`.

```
# Init the result folder structure for the result
result_folder_all = './results'
result_folder = result_folder_all
```

3. Visualization for result

(1) Sample information - PCA plot

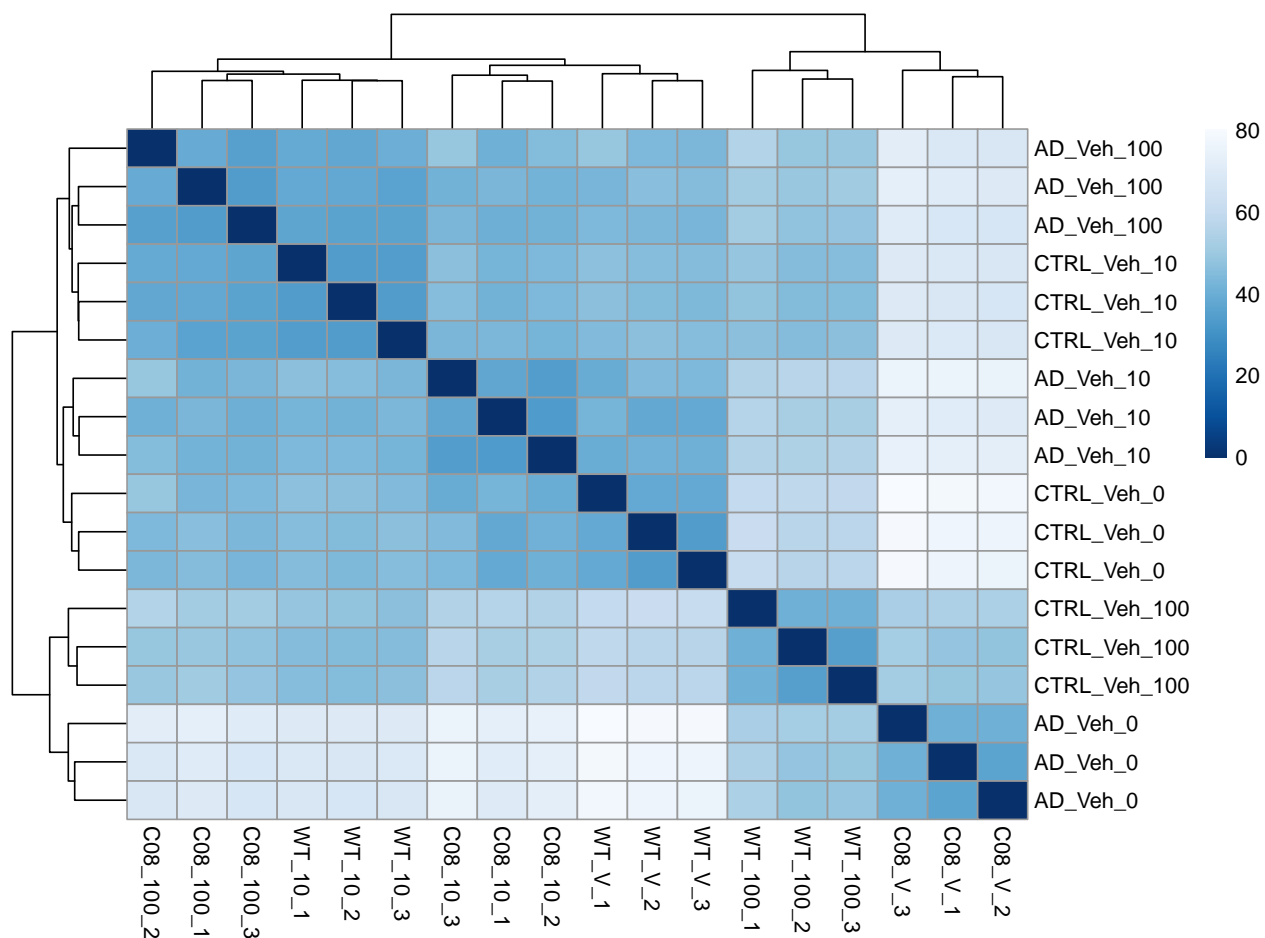
```
figure_folder = result_folder
# do PCA for counts data
dds_obj <- DESeqDataSetFromMatrix(countData = counts,
                                  colData = condition_list,
                                  design = ~ group)
vsd_obj <- varianceStabilizingTransformation(dds_obj, blind = TRUE)
pcaData <- plotPCA(vsd_obj, intgroup = c("group"), returnData = T)
percentVar <- round(100 * attr(pcaData, "percentVar"))

p <- ggplot(pcaData, aes(PC1, PC2, color=group)) +
  geom_point(size=3) +
  labs(x = paste0("PC1: ", percentVar[1], "% variance"),
       y = paste0("PC2: ", percentVar[2], "% variance"),
  ) +
  stat_ellipse(level = 0.95) +
  theme_bw() +
  # theme_classic() +
  theme(text = element_text(family = "Arial", colour = "black")) +
  # scale_color_manual(values = assigned_colors) +
  ggrepel::geom_text_repel(aes(label = name), color = "black")

ggsave("./results/01-Sample_info/01_sample_PCA_plot.pdf", p, width = 8, height = 6, units = "in", dpi = 300)
ggsave("./results/01-Sample_info/01_sample_PCA_plot.png", p, width = 8, height = 6, units = "in", dpi = 300)
```

(2) Sample information - Distance heatmap

```
# Now apply variance stabilizing transformation
sampleDists <- dist(t(assay(vsd_obj)))
sampleDistMatrix <- as.matrix( sampleDists )
rownames(sampleDistMatrix) <- paste( vsd_obj$group )
colors <- colorRampPalette( rev(RColorBrewer::brewer.pal(9, "Blues")) )(255)
p <- pheatmap::pheatmap(sampleDistMatrix,
                        clustering_distance_rows = sampleDists,
                        clustering_distance_cols = sampleDists,
                        col = colors)
```



```
ggsave("./results/01-Sample_info/02_sample_distance_heatmap.pdf", p,width = 8, height = 6, units = "in")
ggsave("./results/01-Sample_info/02_sample_distance_heatmap.png",
p, width = 8, height = 6, units = "in", dpi = 300)
```

4. GSVA analysis

```
## The following code is used to generate the GSVA matrix , only need to run once
# gmxFile <- here::here("data", "ref", "c5.go.v2023.1.Hs.symbols.gmt")
# go_list <- getGmt(gmxFile)
#
# geneset <- go_list
# dat <- as.matrix(counts)
#
# gsvapar <- gsvaParam(dat, geneset, maxDiff=TRUE)
# gsva_es <- gsva(gsvapar)
# gsva_matrix <- as.data.frame(gsva_es)
#
# # save the result
# write.csv(gsva_matrix, "./results/02-GSVA/01_GSVA_matrix.csv")
#
#
# plot the heatmap for the GSVA result
pathway_list <- read.csv(here::here("data", "ref", "focus-pathway_2024_10_03.csv"))

box_plot_folder<- file.path(result_folder,"04-GSVA","Boxplot")
# create the folder
dir.create(box_plot_folder, showWarnings = FALSE)

gsva_matrix <- read.csv("./results/02-GSVA/01_GSVA_matrix.csv", row.names = 1)
colnames(gsva_matrix) <- sub("^X", "", colnames(gsva_matrix))
condition_list_label <- condition_list
condition_list_label$group <- factor(
  condition_list_label$group,
  # levels = c("CTRL_Veh_0", "CTRL_Veh_10", "CTRL_Veh_100",
  #           "AD_Veh_0", "AD_Veh_10", "AD_Veh_100")
  levels = c("CTRL_Veh_0", "AD_Veh_0",
             "CTRL_Veh_10", "AD_Veh_10",
             "CTRL_Veh_100", "AD_Veh_100")
)

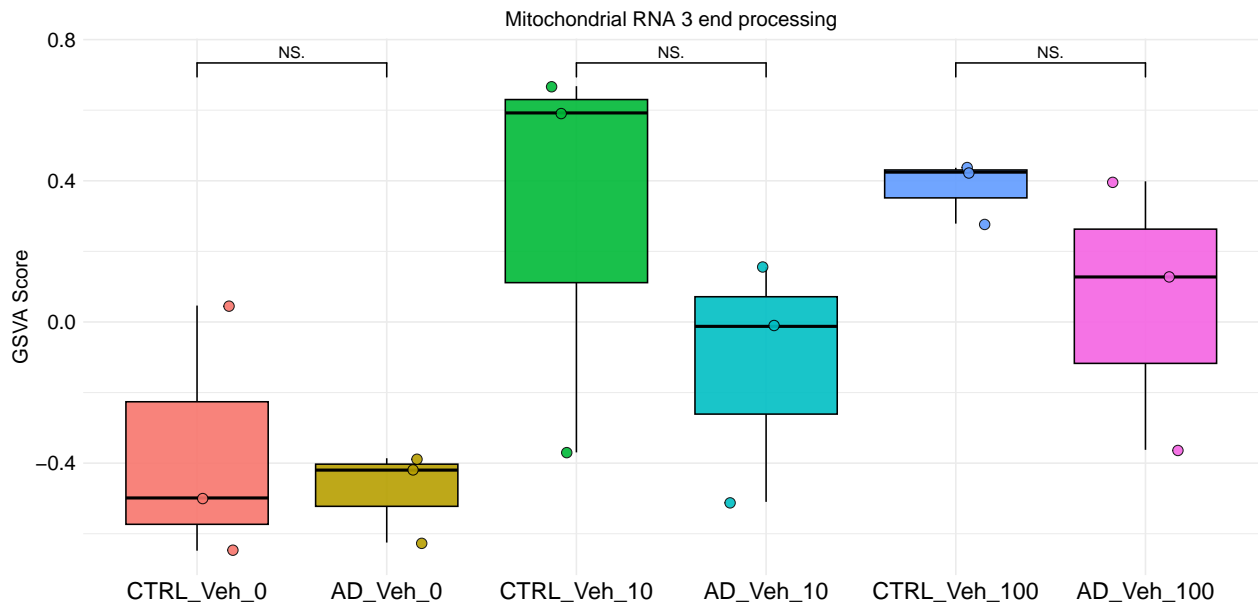
# plot for the focus pathway
for (i in 1:20){
  pathway_name <- pathway_list$pathway[i]
  print(pathway_name)
  p<-plot_gsva_boxplot_mutil_3(gsva_matrix,
    condition_list_label =condition_list_label,
    pathway_name = pathway_name,
    figure_folder = file.path(result_folder,"02-GSVA","Boxplot-pair"),
    file_name = paste0("GSVA_", pathway_name),
    fig.height = 6, fig.width = 12,
    reference_group_1="CTRL_Veh_0" , compare_group_1="AD_Veh_0",
    reference_group_2="CTRL_Veh_10", compare_group_2="AD_Veh_10",
    reference_group_3="CTRL_Veh_100", compare_group_3="AD_Veh_100")

  print(p)
```

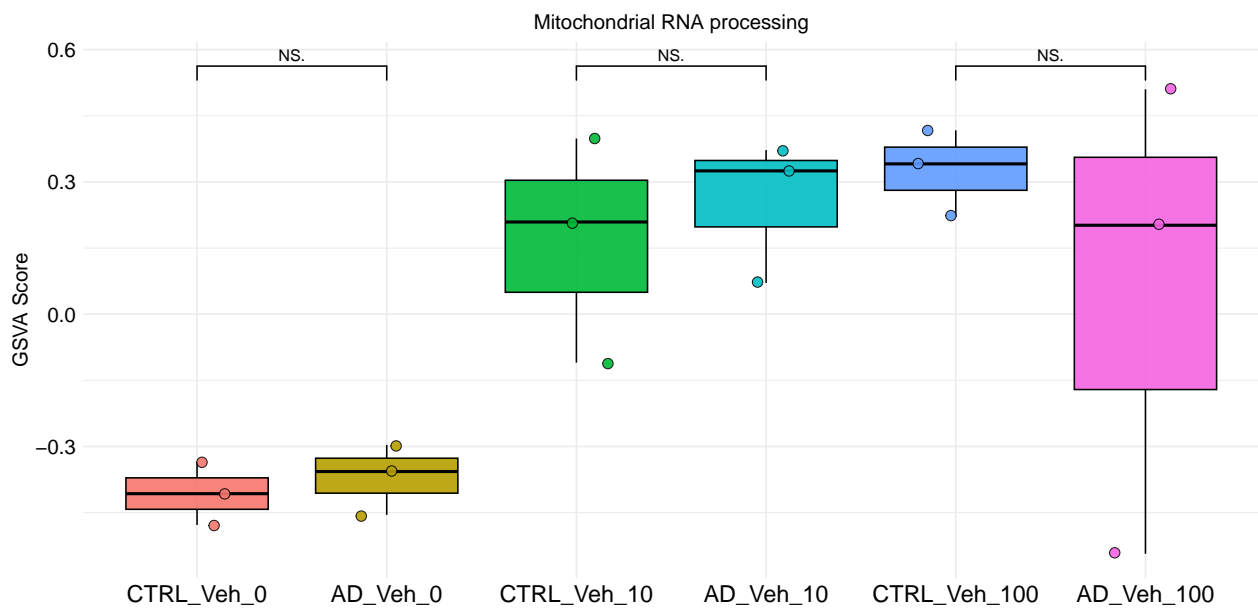
```
}
```

```
## [1] "GOBP_MITOCHONDRIAL_RNA_3_END_PROCESSING"
```

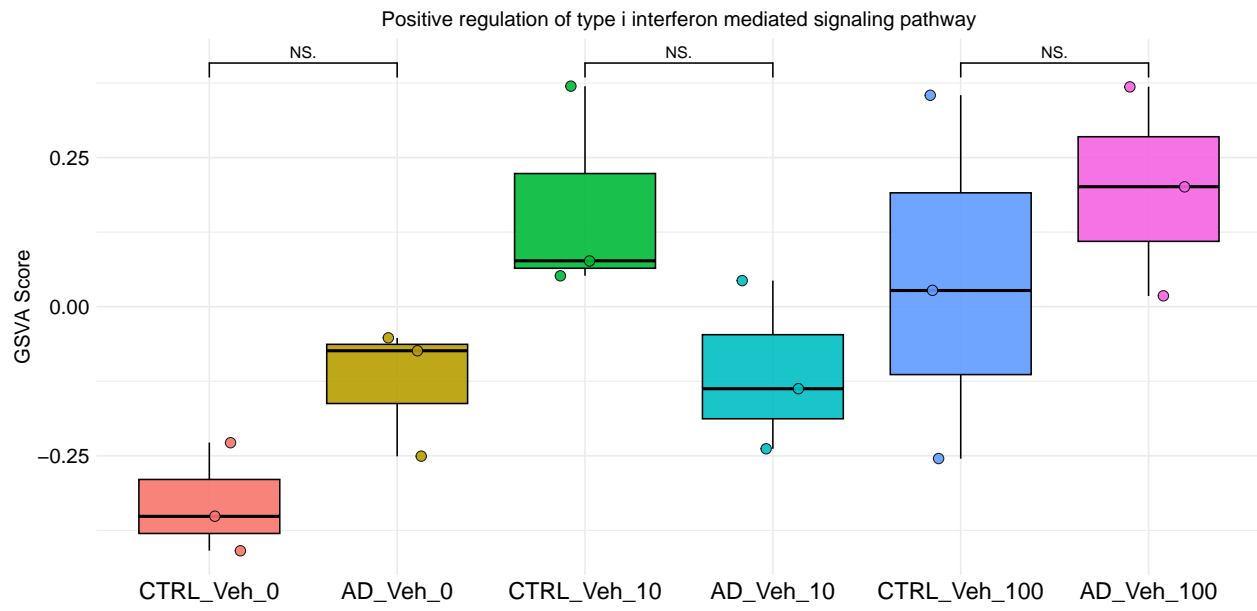
```
## [1] "GOBP_MITOCHONDRIAL_RNA_PROCESSING"
```



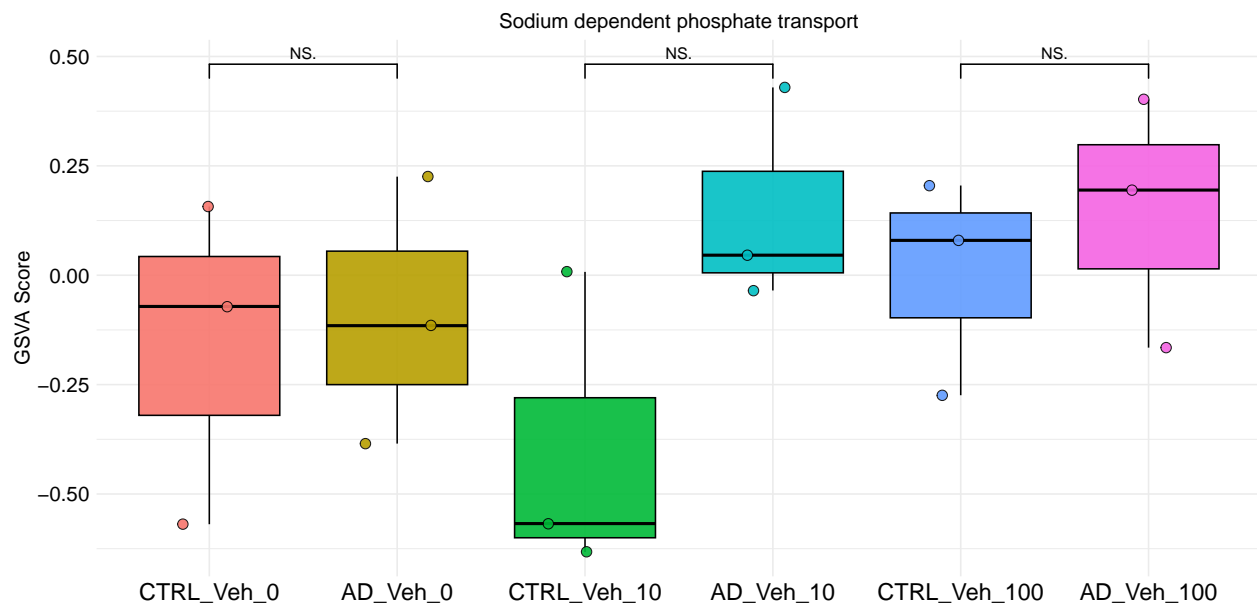
```
## [1] "GOBP_POSITIVE_REGULATION_OF_TYPE_I_INTERFERON_MEDIATED_SIGNALING_PATHWAY"
```



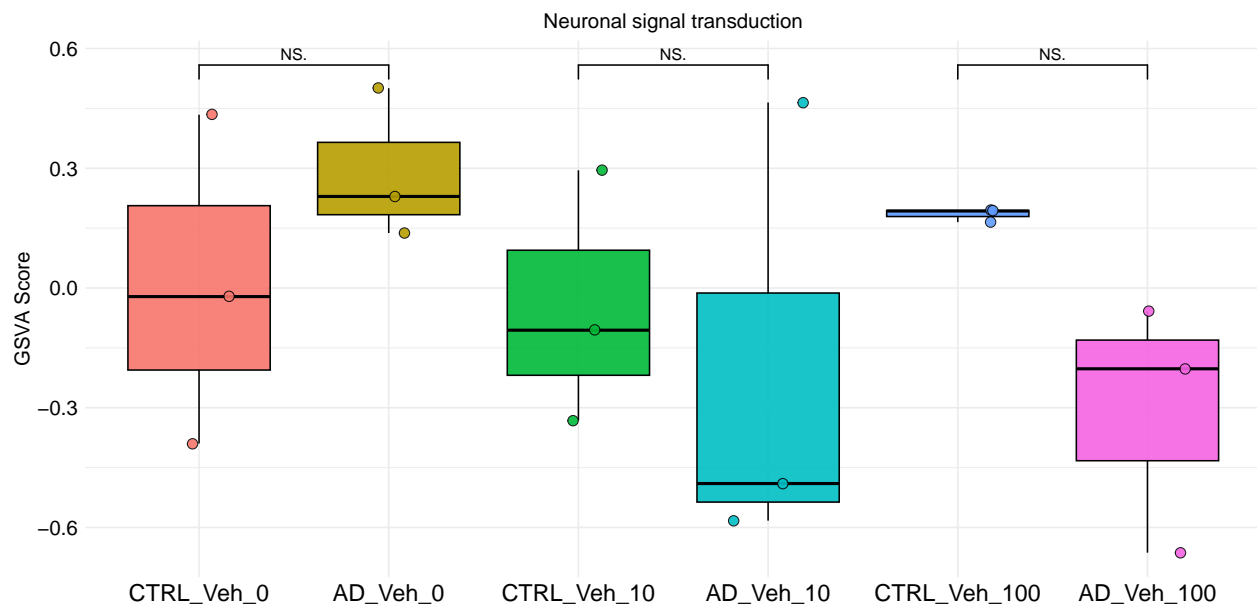
```
## [1] "GOBP_SODIUM_DEPENDENT_PHOSPHATE_TRANSPORT"
```



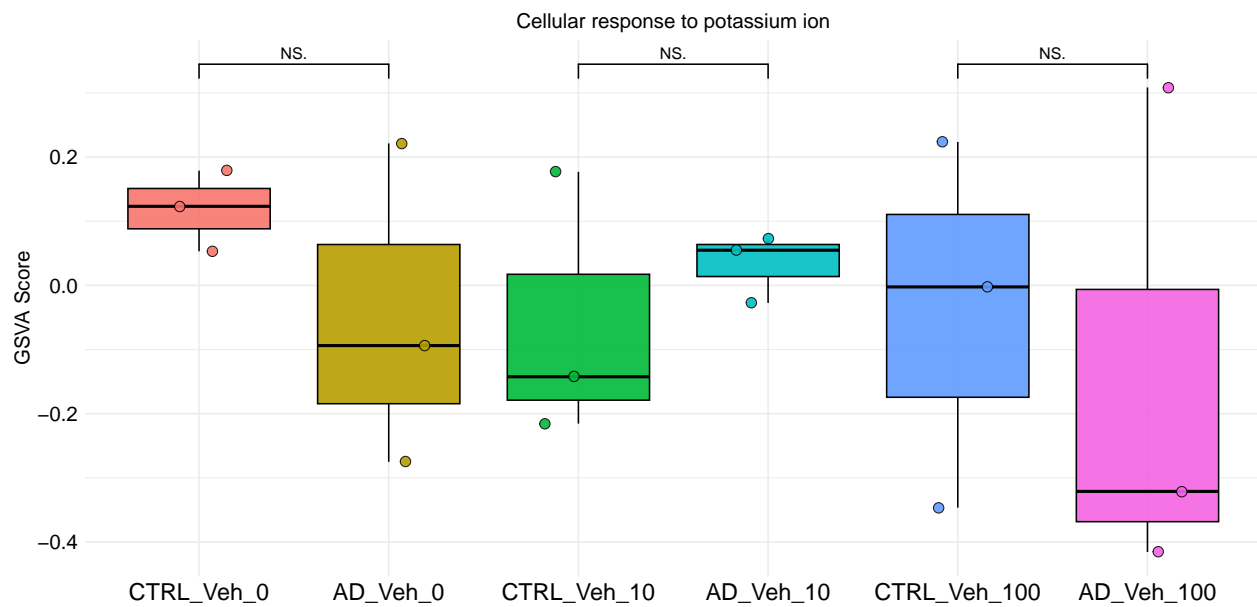
[1] "GOBP_NEURONAL_SIGNAL_TRANSDUCTION"



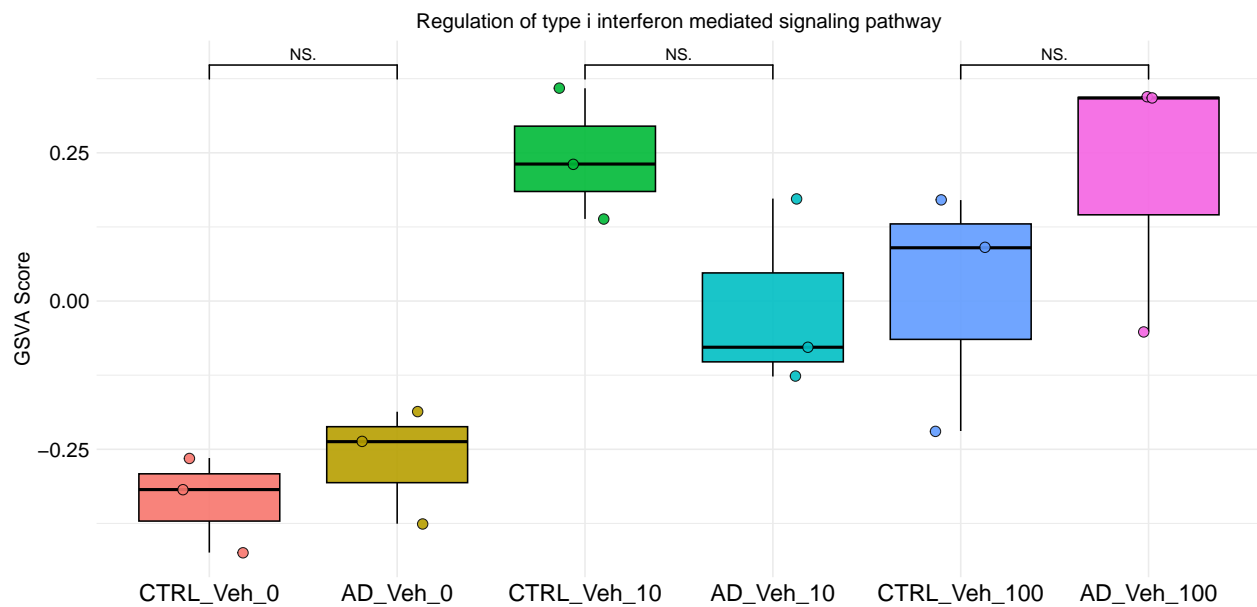
[1] "GOBP_CELLULAR_RESPONSE_TO_POTASSIUM_ION"



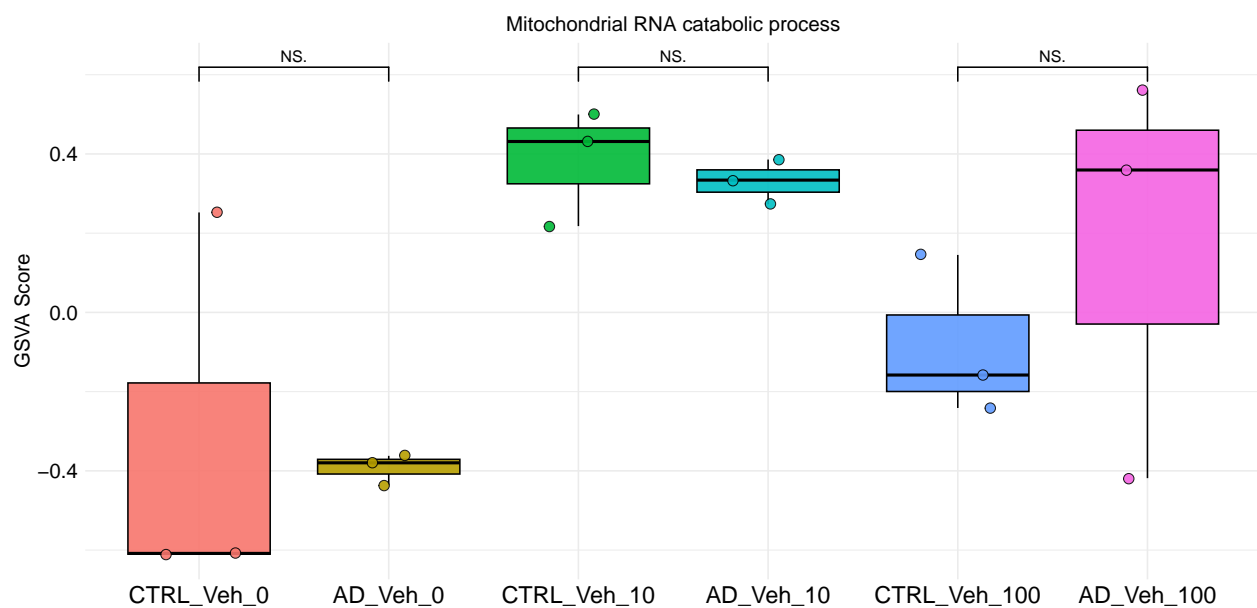
[1] "GOBP_REGULATION_OF_TYPE_I_INTERFERON_MEDIATED_SIGNALING_PATHWAY"



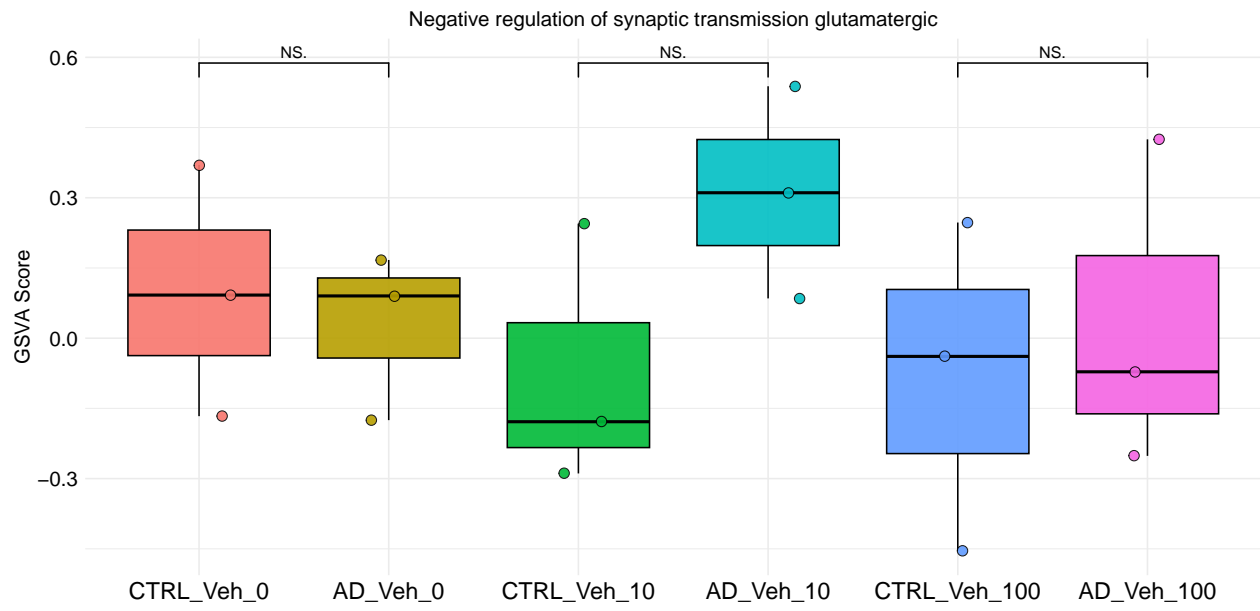
[1] "GOBP_MITOCHONDRIAL_RNA_CATABOLIC_PROCESS"



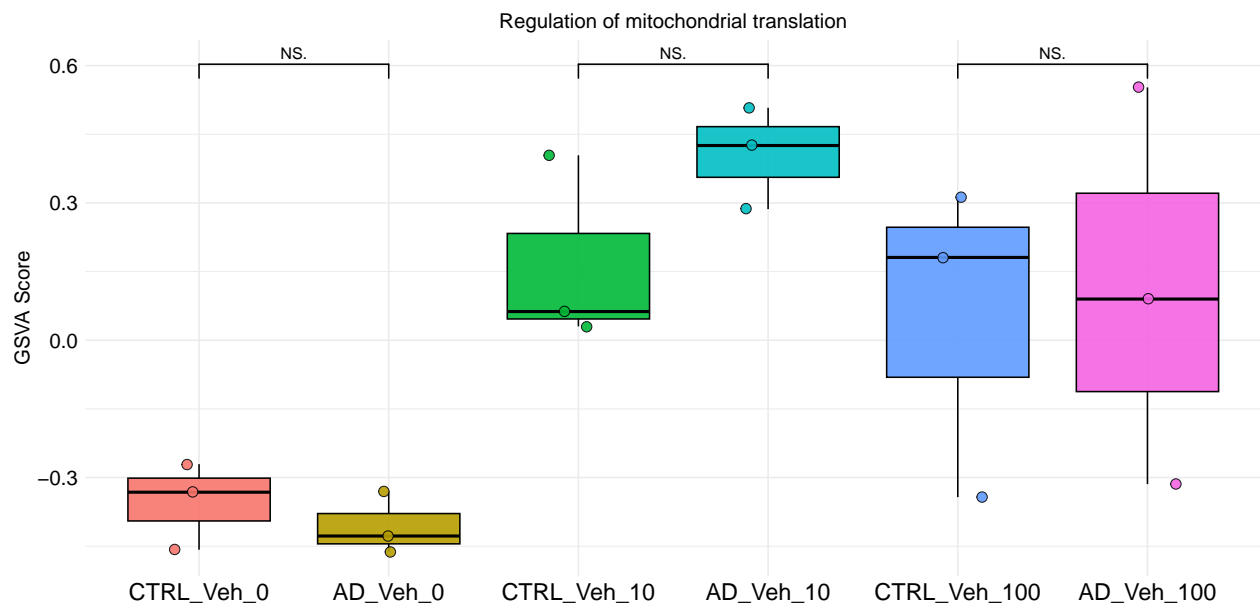
[1] "GOBP_NEGATIVE_REGULATION_OF_SYNAPTIC_TRANSMISSION_Glutamatergic"



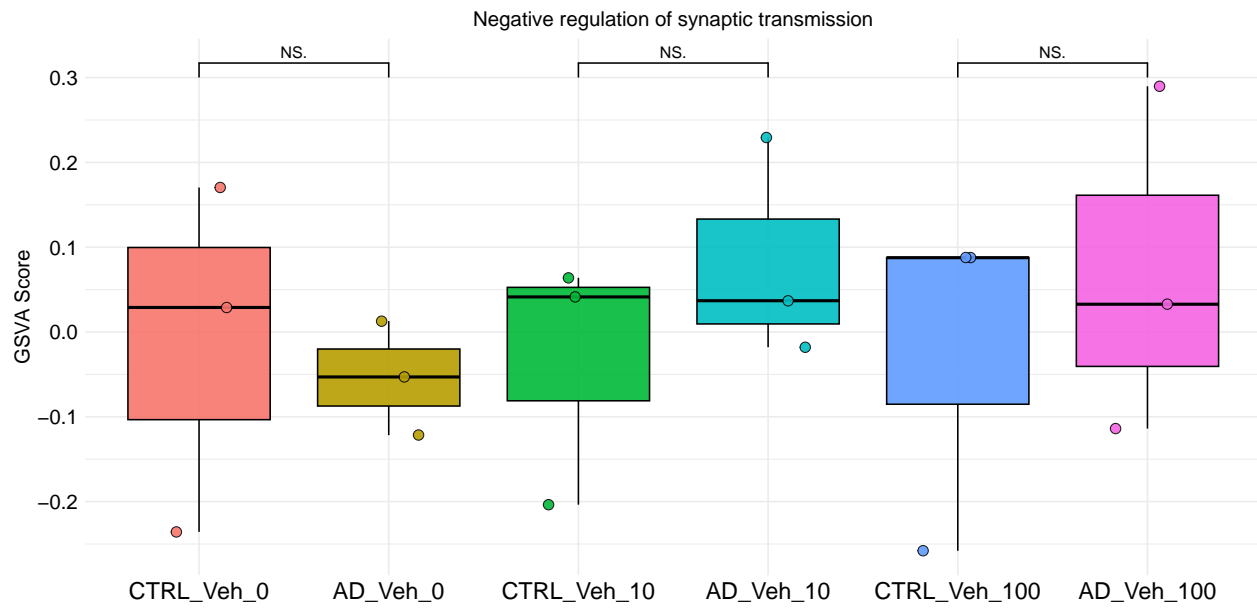
[1] "GOBP_REGULATION_OF_MITOCHONDRIAL_TRANSLATION"



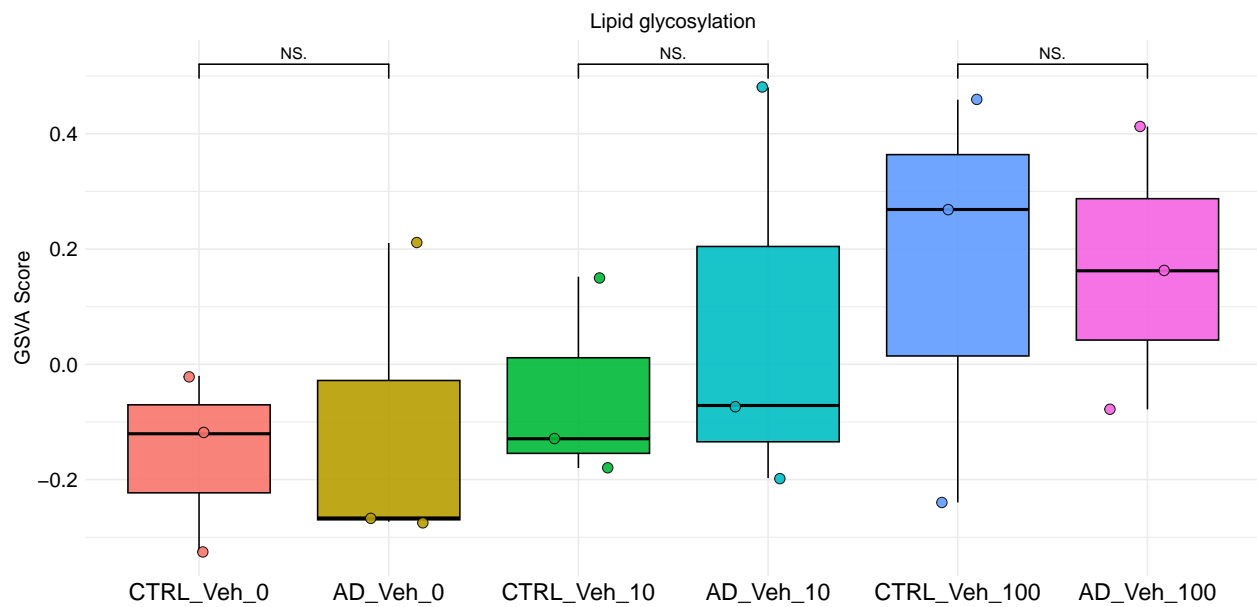
[1] "GOBP_NEGATIVE_REGULATION_OF_SYNAPTIC_TRANSMISSION"



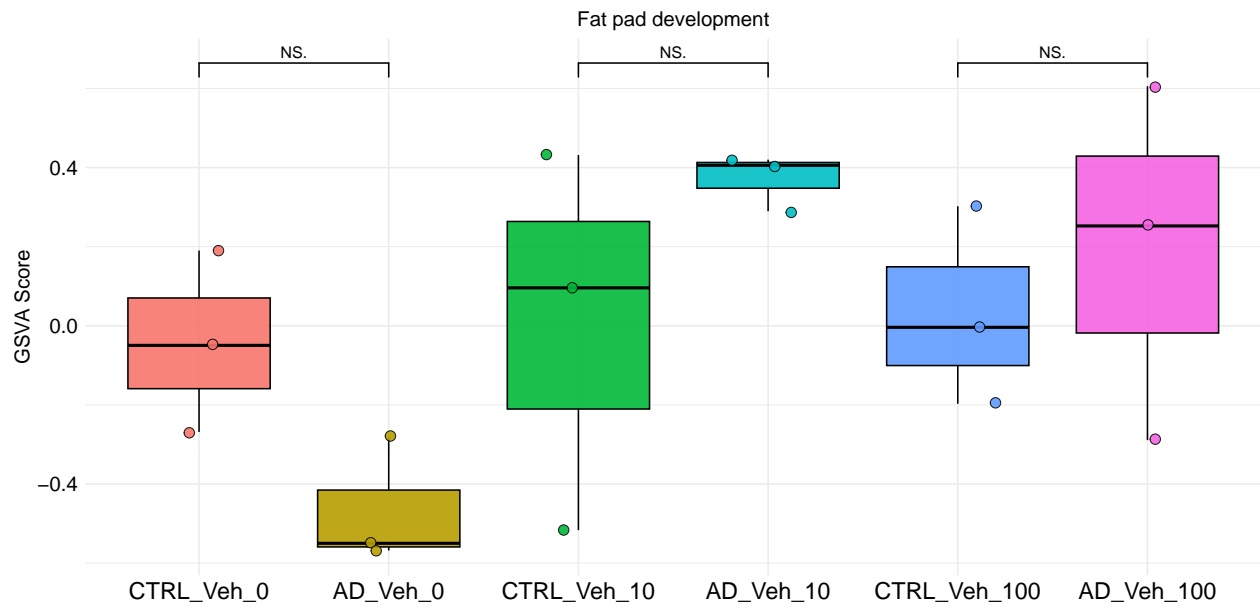
[1] "GOBP_LIPID_GLYCOSYLATION"



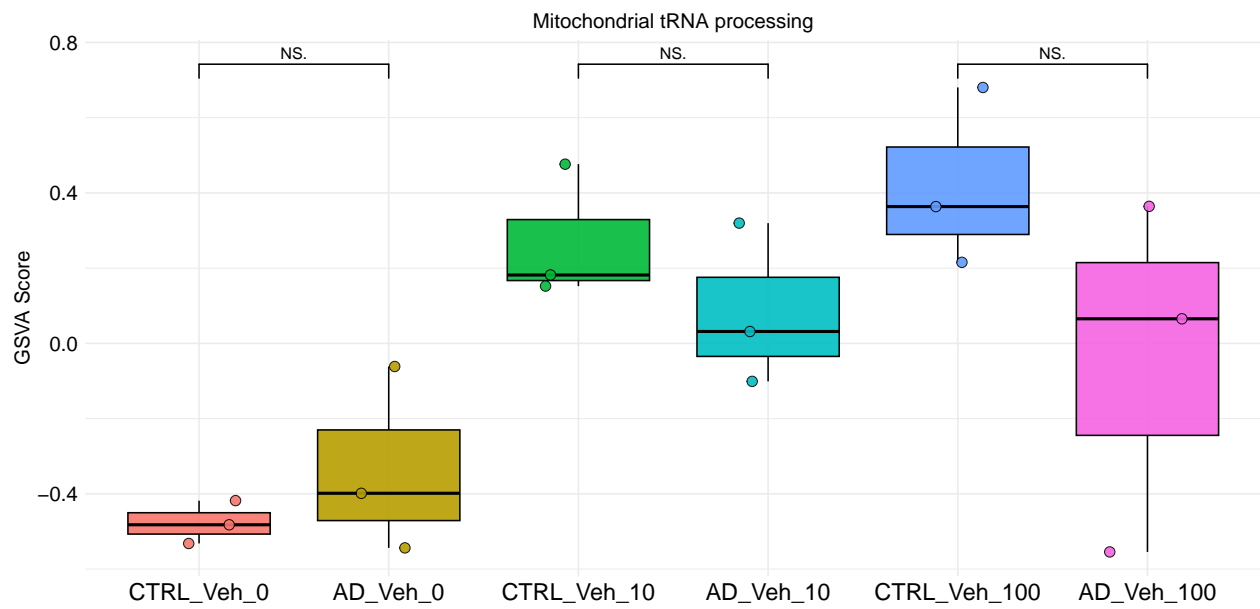
[1] "GOBP_FAT_PAD_DEVELOPMENT"



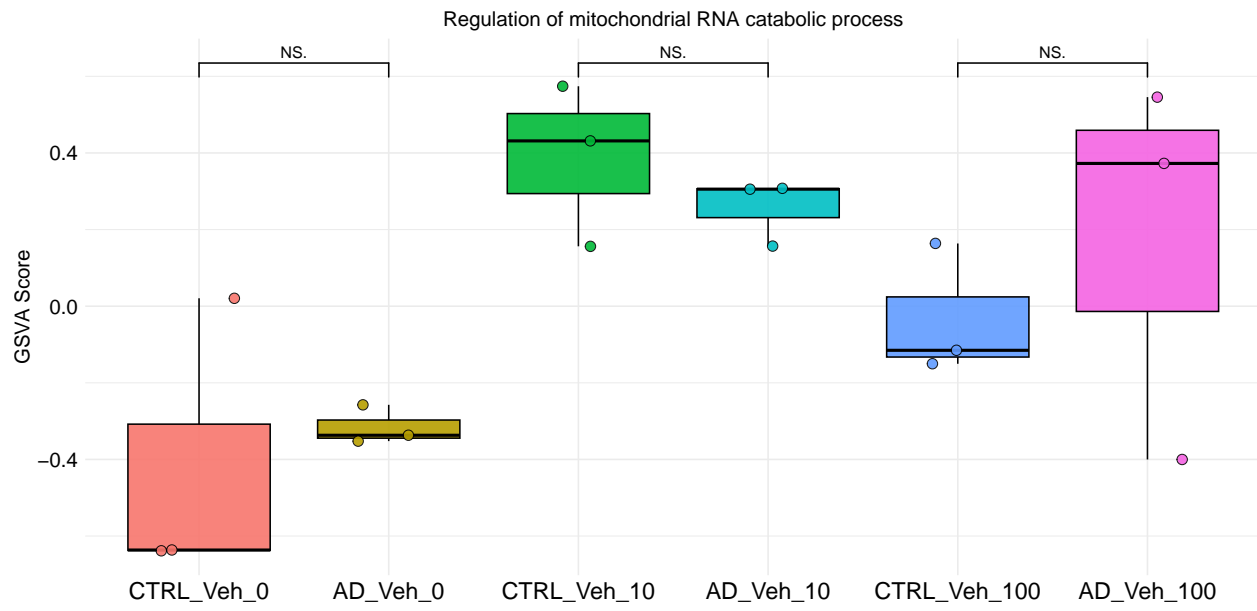
[1] "GOBP_MITOCHONDRIAL_TRNA_PROCESSING"



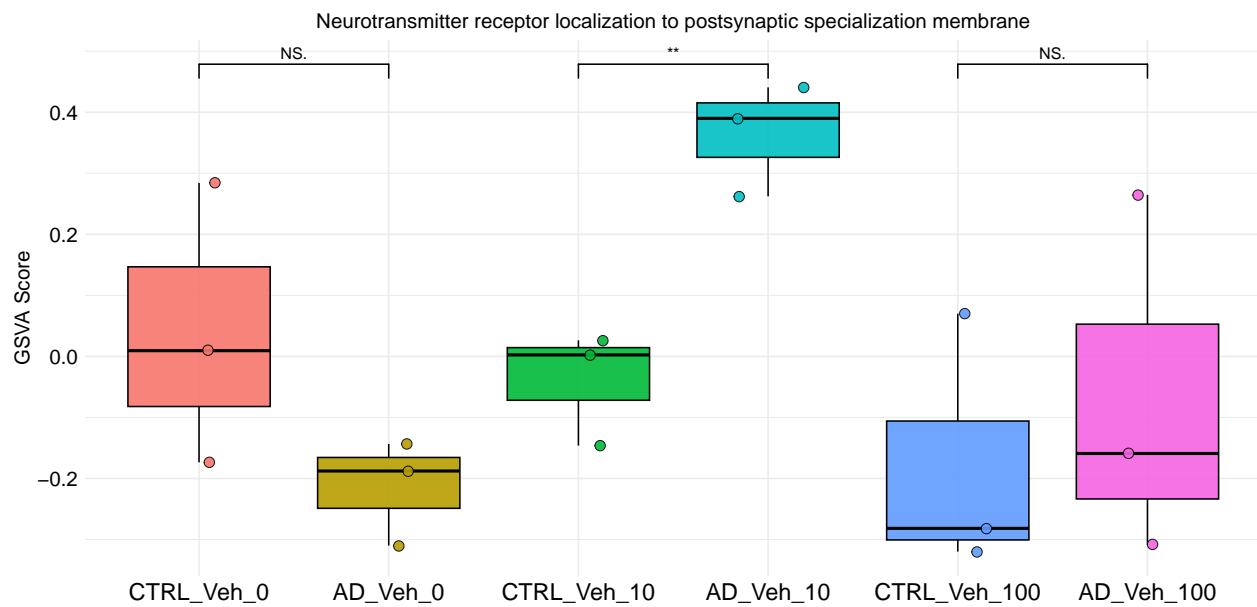
[1] "GOBP_REGULATION_OF_MITOCHONDRIAL_RNA_CATABOLIC_PROCESS"



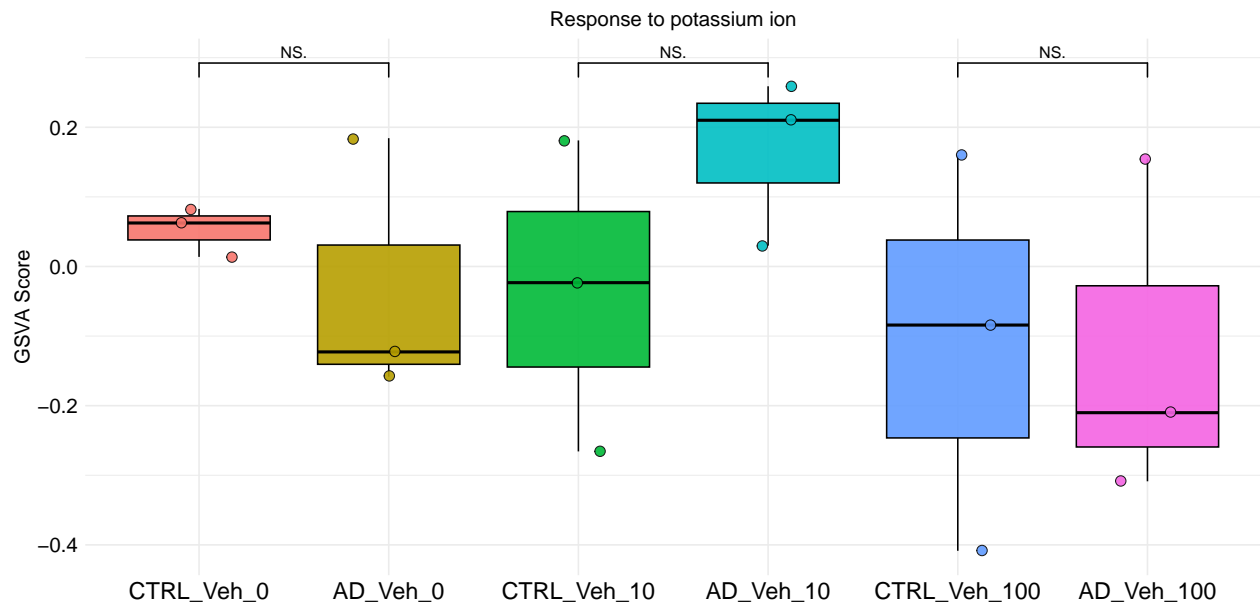
[1] "GOBP_NEUROTRANSMITTER_RECEPTOR_LOCALIZATION_TO_POSTSYNAPTIC_SPECIALIZATION_MEMBRANE"



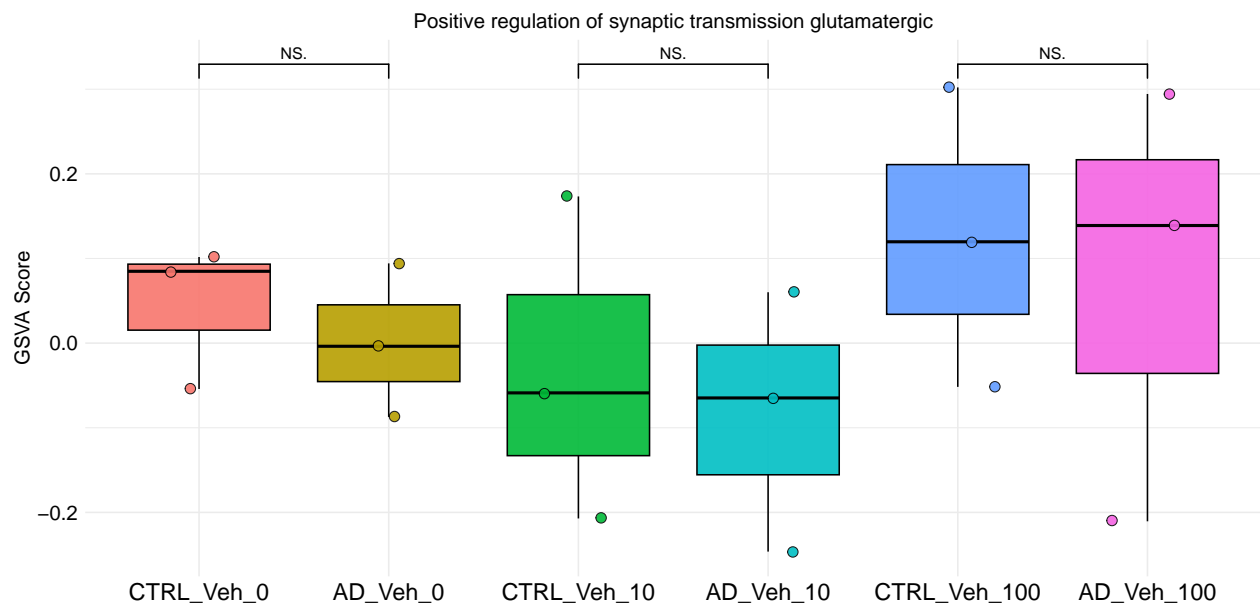
[1] "GOBP_RESPONSE_TO_POTASSIUM_ION"



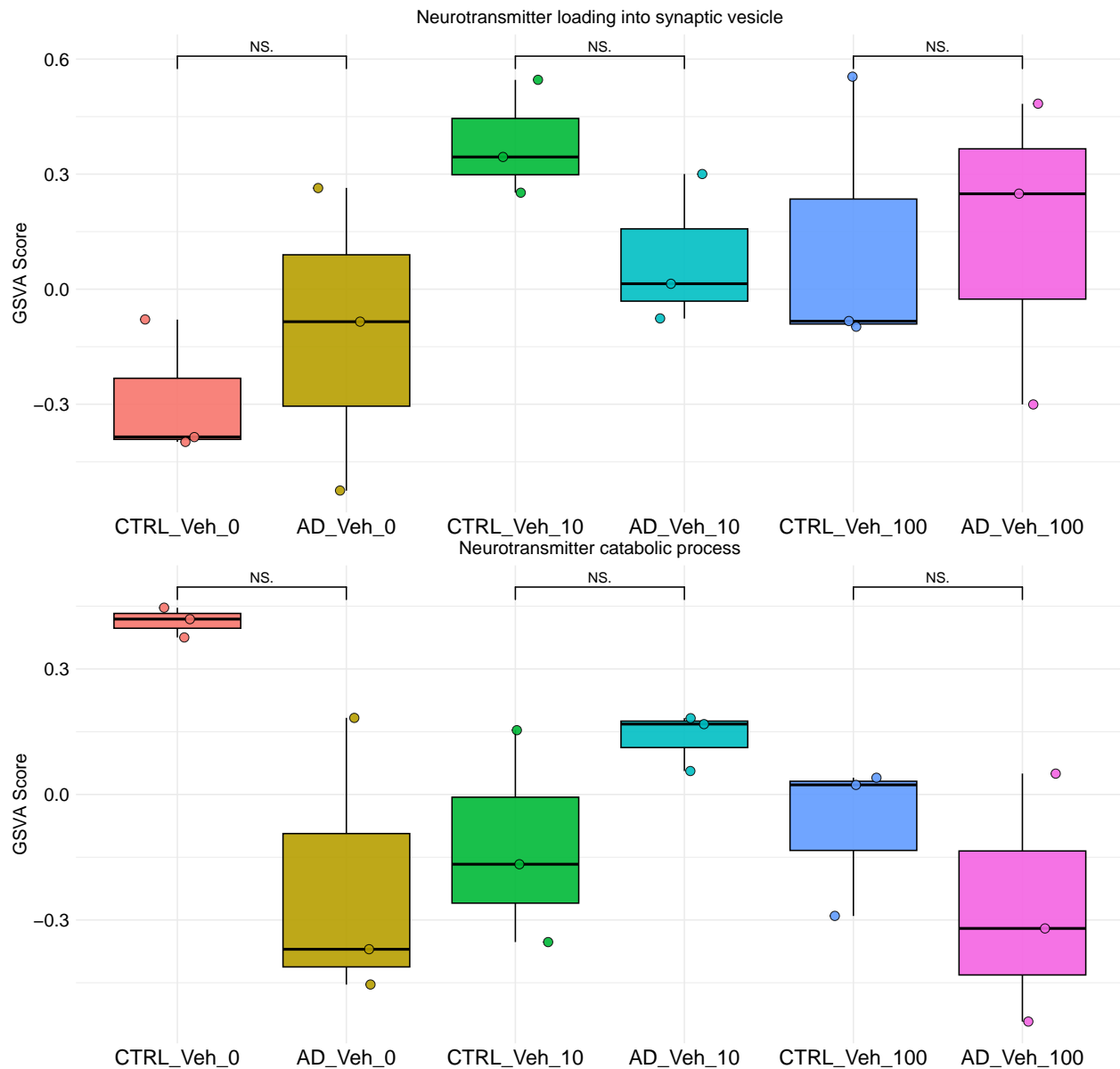
[1] "GOBP_POSITIVE_REGULATION_OF_SYNAPTIC_TRANSMISSION_GLUTAMATERGIC"



[1] "GOBP_NEUROTRANSMITTER_LOADING_INTO_SYNAPTIC_VESICLE"



[1] "GOBP_NEUROTRANSMITTER_CATABOLIC_PROCESS"



Session information

```
sessionInfo()
```

```
## R version 4.4.0 (2024-04-24)
## Platform: aarch64-apple-darwin20
## Running under: macOS Sonoma 14.3.1
##
## Matrix products: default
## BLAS:   /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRlapack.dylib; LAPACK v
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
```

```

## time zone: America/New_York
## tzcode source: internal
##
## attached base packages:
## [1] parallel stats4 stats graphics grDevices utils datasets
## [8] methods base
##
## other attached packages:
## [1] GSEABase_1.66.0 graph_1.82.0
## [3] annotate_1.82.0 XML_3.99-0.18
## [5] extrafont_0.19 ggsignif_0.6.4
## [7] patchwork_1.3.0 decoupleR_2.10.0
## [9] GSVA_1.52.3 BiocParallel_1.38.0
## [11] edgeR_4.2.2 limma_3.60.6
## [13] GenomicFeatures_1.56.0 biomaRt_2.60.1
## [15] gprofiler2_0.2.3 RColorBrewer_1.1-3
## [17] data.table_1.16.4 org.Hs.eg.db_3.19.1
## [19] AnnotationDbi_1.66.0 clusterProfiler_4.12.6
## [21] ggfortify_0.4.17 pheatmap_1.0.12
## [23] EnhancedVolcano_1.22.0 ggrepel_0.9.6
## [25] apeglm_1.26.1 DESeq2_1.44.0
## [27] SummarizedExperiment_1.34.0 Biobase_2.64.0
## [29] MatrixGenerics_1.16.0 matrixStats_1.5.0
## [31] reshape2_1.4.4 Matrix_1.7-2
## [33] Signac_1.14.0 Seurat_5.2.1
## [35] SeuratObject_5.0.2 sp_2.2-0
## [37] rtracklayer_1.64.0 GenomicRanges_1.56.2
## [39] GenomeInfoDb_1.40.1 IRanges_2.38.1
## [41] S4Vectors_0.42.1 BiocGenerics_0.50.0
## [43] knitr_1.49 lubridate_1.9.4
## [45] forcats_1.0.0 stringr_1.5.1
## [47] dplyr_1.1.4 purrr_1.0.4
## [49] readr_2.1.5 tidyr_1.3.1
## [51] tibble_3.2.1 ggplot2_3.5.1
## [53] tidyverse_2.0.0
##
## loaded via a namespace (and not attached):
## [1] SpatialExperiment_1.14.0 R.methodsS3_1.8.2
## [3] progress_1.2.3 goftest_1.2-3
## [5] HDF5Array_1.32.1 Biostrings_2.72.1
## [7] vctrs_0.6.5 spatstat.random_3.3-2
## [9] digest_0.6.37 png_0.1-8
## [11] deldir_2.0-4 parallelly_1.42.0
## [13] magick_2.8.5 MASS_7.3-64
## [15] httpuv_1.6.15 qvalue_2.36.0
## [17] withr_3.0.2 xfun_0.51
## [19] ggfun_0.1.8 survival_3.8-3
## [21] memoise_2.0.1 gson_0.1.0
## [23] systemfonts_1.2.1 ragg_1.3.3
## [25] tidytree_0.4.6 zoo_1.8-12
## [27] pbapply_1.7-2 R.oo_1.27.0
## [29] prettyunits_1.2.0 KEGGREST_1.44.1
## [31] promises_1.3.2 httr_1.4.7
## [33] restfulr_0.0.15 rhdf5filters_1.16.0

```


## [35]	globals_0.16.3	fitdistrplus_1.2-2
## [37]	rhdf5_2.48.0	rstudioapi_0.17.1
## [39]	UCSC.utils_1.0.0	miniUI_0.1.1.1
## [41]	generics_0.1.3	DOSE_3.30.5
## [43]	curl_6.2.1	zlibbioc_1.50.0
## [45]	ScaledMatrix_1.12.0	ggraph_2.2.1
## [47]	polyclip_1.10-7	GenomeInfoDbData_1.2.12
## [49]	SparseArray_1.4.8	xtable_1.8-4
## [51]	evaluate_1.0.3	S4Arrays_1.4.1
## [53]	BiocFileCache_2.12.0	hms_1.1.3
## [55]	irlba_2.3.5.1	colorspace_2.1-1
## [57]	filelock_1.0.3	ROCR_1.0-11
## [59]	reticulate_1.40.0	spatstat.data_3.1-4
## [61]	magrittr_2.0.3	lmtest_0.9-40
## [63]	later_1.4.1	viridis_0.6.5
## [65]	ggtree_3.12.0	lattice_0.22-6
## [67]	spatstat.geom_3.3-5	future.apply_1.11.3
## [69]	scattermore_1.2	shadowtext_0.1.4
## [71]	cowplot_1.1.3	RcppAnnoy_0.0.22
## [73]	pillar_1.10.1	nlme_3.1-167
## [75]	compiler_4.4.0	beachmat_2.20.0
## [77]	RSpectra_0.16-2	stringi_1.8.4
## [79]	tensor_1.5	GenomicAlignments_1.40.0
## [81]	plyr_1.8.9	crayon_1.5.3
## [83]	abind_1.4-8	BiocIO_1.14.0
## [85]	gridGraphics_0.5-1	emdbbook_1.3.13
## [87]	locfit_1.5-9.11	graphlayouts_1.2.2
## [89]	bit_4.5.0.1	fastmatch_1.1-6
## [91]	textshaping_1.0.0	codetools_0.2-20
## [93]	BiocSingular_1.20.0	plotly_4.10.4
## [95]	mime_0.12	splines_4.4.0
## [97]	Rcpp_1.0.14	fastDummies_1.7.5
## [99]	sparseMatrixStats_1.16.0	dbplyr_2.5.0
## [101]	Rttf2pt1_1.3.12	blob_1.2.4
## [103]	here_1.0.1	fs_1.6.5
## [105]	listenv_0.9.1	ggplotify_0.1.2
## [107]	statmod_1.5.0	tzdb_0.4.0
## [109]	tweenr_2.0.3	pkgconfig_2.0.3
## [111]	tools_4.4.0	cachem_1.1.0
## [113]	RSQLite_2.3.9	viridisLite_0.4.2
## [115]	DBI_1.2.3	numDeriv_2016.8-1.1
## [117]	fastmap_1.2.0	rmarkdown_2.29
## [119]	scales_1.3.0	grid_4.4.0
## [121]	ica_1.0-3	Rsamtools_2.20.0
## [123]	coda_0.19-4.1	dotCall64_1.2
## [125]	RANN_2.6.2	farver_2.1.2
## [127]	tidygraph_1.3.1	scatterpie_0.2.4
## [129]	yaml_2.3.10	cli_3.6.4
## [131]	lifecycle_1.0.4	uwot_0.2.2
## [133]	mvtnorm_1.3-3	timechange_0.3.0
## [135]	gtable_0.3.6	rjson_0.2.23
## [137]	ggribges_0.5.6	progressr_0.15.1
## [139]	ape_5.8-1	jsonlite_1.9.0
## [141]	RcppHNSW_0.6.0	bitops_1.0-9

## [143] bit64_4.6.0-1	Rtsne_0.17
## [145] yulab.utils_0.2.0	spatstat.utils_3.1-2
## [147] bdsmatrix_1.3-7	GOSemSim_2.30.2
## [149] spatstat.univar_3.1-1	R.utils_2.12.3
## [151] lazyeval_0.2.2	shiny_1.10.0
## [153] htmltools_0.5.8.1	enrichplot_1.24.4
## [155] GO.db_3.19.1	sctransform_0.4.1
## [157] rappdirs_0.3.3	tinytex_0.55
## [159] glue_1.8.0	spam_2.11-1
## [161] httr2_1.1.0	XVector_0.44.0
## [163] RCurl_1.98-1.16	rprojroot_2.0.4
## [165] treeio_1.28.0	gridExtra_2.3
## [167] extrafontdb_1.0	igraph_2.1.4
## [169] R6_2.6.1	SingleCellExperiment_1.26.0
## [171] labeling_0.4.3	RcppRoll_0.3.1
## [173] cluster_2.1.8	bbmle_1.0.25.1
## [175] Rhdf5lib_1.26.0	aplot_0.2.4
## [177] DelayedArray_0.30.1	tidyselect_1.2.1
## [179] ggforce_0.4.2	xml2_1.3.6
## [181] future_1.34.0	rsvd_1.0.5
## [183] munsell_0.5.1	KernSmooth_2.23-26
## [185] htmlwidgets_1.6.4	fgsea_1.30.0
## [187] rlang_1.1.5	spatstat.sparse_3.1-0
## [189] spatstat.explore_3.3-4	