

eCB expression analysis of Hypothalamus development with focus on PVN

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Setup parameters

```
# Load tidyverse infrastructure packages
suppressPackageStartupMessages({
  library(future)
  library(here)
  library(tidyverse)
  library(magrittr)
  library(stringr)
  library(skimr)
  library(RColorBrewer)
  library(viridis)
})

# Load packages for scRNA-seq analysis and visualisation
suppressPackageStartupMessages({
  library(UpSetR)
  library(ggplot2)
  library(cowplot)
  library(patchwork)
  library(ggstatsplot)
  library(anndata)
```

```

library(sceasy)
library(Seurat)
library(SeuratDisk)
library(SeuratWrappers)
library(scCustomize)
})

sc <- import("scanpy", convert = FALSE)

```

Set paths

```

src_dir <- here("code")
data_dir <- here("data")
output_dir <- here("output")
plots_dir <- here(output_dir, "figures/")
tables_dir <- here(output_dir, "tables/")

```

Load helper functions and gene-sets

```

source(here(src_dir, "genes.R"))
source(here(src_dir, "functions.R"))

```

Set fixed variables

```

# set seed
reseed <- 42
set.seed(seed = reseed)

# Parameters for parallel execution
n_cores <- 8
plan("multisession", workers = n_cores)
options(
  future.globals.maxSize = 100000 * 1024^2,
  future.rng.onMisuse = "ignore"
)

```



```
plan()
```

```
multisession:
```

```
- args: function (... , workers = 8, envir = parent.frame())  
- tweaked: TRUE  
- call: plan("multisession", workers = n_cores)
```

```
# ggplot2 theme  
theme_set(ggmin::theme_powerpoint())
```

```
bioproject <- "PRJNA547712"  
project <- "kim2020_Hypoth-dev"  
cb_fpr <- 0.001  
low_cutoff_gene <- 500  
high_cutoff_gene <- NULL  
high_cutoff_gene <- 5000  
low_cutoff_umis <- NULL  
low_cutoff_umis <- -Inf  
high_cutoff_umis <- 25000  
high_cutoff_pc_mt <- 15  
high_cutoff_pc_ribo <- 20  
high_cutoff_pc_hb <- 0.1  
high_cutoff_doublet_score <- 0.33  
high_cutoff_complexity <- 0.85  
connectivity_model <- "min_tree"  
k <- 10  
metric <- "euclidean"  
signature <- 100
```

Load Kim DW et al 2020

```
anndata <- sc$read(here(  
  "kim2020_combined.h5ad"  
))
```

Convert adata object to R AnnDataR6 object.

```
adata <- py_to_r(anndata)
class(adata)
```

```
[1] "AnnDataR6" "R6"
```

```
class(adata$X)
```

```
[1] "dgRMatrix"
attr(,"package")
[1] "Matrix"
```

```
adata
```

AnnData object with n_obs × n_vars = 128006 × 27998

```
obs: 'bc_name', 'orig.ident', 'nCount_RNA', 'nFeature_RNA', 'X1', 'X', 'Y', 'Z', '
var: '_index', 'features'
```

```
srt_path <- here(
  "data",
  sprintf("%s-whole_dataset-simple.h5Seurat", bioproject)
)
```

```
expr_mtx <- t(as.matrix(adata$X))
colnames(expr_mtx) <- rownames(adata$X)
rownames(expr_mtx) <- adata$var$features
srt <- CreateSeuratObject(
  expr_mtx,
  assay = "RNA",
  project = "kim2020_Hypoth_dev",
  meta.data = as.data.frame(adata$obs)
)
```

```
X_umap <- adata$obs |>
  select(X, Y) |>
```

```

    as.matrix()
colnames(X_umap) <- c("UMAP_1", "UMAP_2")
rownames(X_umap) <- colnames(expr_mtx)
srt[["umap"]] <- CreateDimReducObject(embeddings = X_umap, key = "umap_", assay = De

Idents(srt) <- "age"
srt <- Store_Palette_Seurat(seurat_object = srt, palette = rev(brewer.pal(n = 11, na

```

Load Romanov et al 2020

```
print(srt)
```

An object of class Seurat
 27998 features across 128006 samples within 1 assay
 Active assay: RNA (27998 features, 0 variable features)
 1 layer present: counts
 1 dimensional reduction calculated: umap

```

rar2020.srt.pub <- readRDS("/data/1_heteroAstrocytes/PRJNA548917/old/oldCCA_nae_srt.
rar2020.srt.pub <- UpdateSeuratObject(rar2020.srt.pub)
Idents(rar2020.srt.pub) <-
  factor(rar2020.srt.pub$wtree,
    ordered = TRUE
  )

# Consistent colours and clusters names
colours_wtree <- setNames(read_lines(here(data_dir, "colours_wtree.tsv")), 1:45)

rar2020.srt.pub$age <-
  Cells(rar2020.srt.pub) |>
  str_split(pattern = ":", simplify = T) %>%
  .[, 1] %>%
  str_split_fixed(pattern = "_", n = 3) %>%
  .[, 3]
print(rar2020.srt.pub)

```

An object of class Seurat
 24340 features across 51199 samples within 1 assay
 Active assay: RNA (24340 features, 3500 variable features)
 3 layers present: counts, data, scale.data
 3 dimensional reductions calculated: pca, tsne, umap

```
glimpse(rar2020.srt.pub@meta.data)
```

Rows: 51,199

Columns: 20

```
$ nGene          <int> 1652, 782, 447, 1706, 1106, 894, 727, 734, 669, 617, ~
$ nUMI          <dbl> 2787, 1090, 544, 2709, 1817, 1220, 995, 1036, 920, 86~
$ orig.ident    <fct> Hypothalamus, Hypothalamus, Hypothalamus, Hypothalamu~
$ res.0.2       <chr> "23", "23", "23", "23", "23", "23", "23", "23", "23", ~
$ res.0.4       <chr> "34", "34", "34", "34", "34", "34", "34", "34", "34", ~
$ res.0.8       <chr> "42", "42", "42", "42", "42", "42", "42", "42", "42", ~
$ res.1.2       <chr> "47", "47", "47", "47", "47", "47", "47", "47", "47", ~
$ res.2         <chr> "54", "54", "54", "54", "54", "54", "54", "54", "54", ~
$ tree.ident    <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, ~
$ pro_Inter     <chr> "41", "41", "41", "41", "41", "41", "41", "41", "41", ~
$ pro_Enter     <chr> "41", "41", "41", "41", "41", "41", "41", "41", "41", ~
$ tree_final    <fct> 19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 1~
$ subtree       <fct> 41, 41, 41, 41, 41, 41, 41, 41, 41, 41, 41, 41, 41, 4~
$ prim_walktrap <fct> 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 3~
$ umi_per_gene  <dbl> 1.687046, 1.393862, 1.217002, 1.587925, 1.642857, 1.3~
$ log_umi_per_gene <dbl> 0.22712693, 0.14421974, 0.08529138, 0.20082998, 0.215~
$ nCount_RNA    <dbl> 2787, 1090, 544, 2709, 1817, 1220, 995, 1036, 920, 86~
$ nFeature_RNA  <int> 1652, 782, 447, 1706, 1106, 894, 727, 734, 669, 617, ~
$ wtree         <fct> 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 3~
$ age          <chr> "P23", "3P2", "3P2", "P2", "P2", "P2", "P2", "P2", "P~
```

```
table(Idsents(rar2020.srt.pub))
```

1	2	3	4	5	6	7	8	9	10	11	12	13
2344	8146	395	402	3234	712	552	374	259	952	13727	1615	765
14	15	16	17	18	19	20	21	22	23	24	25	26
832	1244	792	590	808	2486	1683	628	1039	1750	292	394	547

27	28	29	30	31	32	33	34	35	36	37	38	39
391	407	507	93	81	402	143	701	222	353	324	73	78
40	41	42	43	44	45							
328	190	73	37	179	55							

```
rar2020.srt.pub %<>% RenameIds(object = ., `43` = "mneOXY")
rar2020.srt.pub %<>% RenameIds(object = ., `26` = "mneVAS")
rar2020.srt.pub %<>% RenameIds(object = ., `31` = "pneSS")
rar2020.srt.pub %<>% RenameIds(object = ., `24` = "pneCRH")
rar2020.srt.pub %<>% RenameIds(object = ., `15` = "pneTRH")
```

```
rar2020.srt.pub$stage <-
  rar2020.srt.pub$age %>%
  forcats::fct_collapse(
    Embryonic = c("E15", "E17"),
    Neonatal = c("P0", "P2", "3P2"),
    Pubertal = c("1P10", "P10"),
    Adult = c("P23")
  )
rar2020.srt.pub$stage %<>% factor(levels = c("Embryonic", "Neonatal", "Pubertal", "Adult"))
rar2020.srt.pub$stage %>% forcats::fct_count()
```

```
# A tibble: 4 x 2
  f      n
<fct> <int>
1 Embryonic 19503
2 Neonatal 20316
3 Pubertal 8965
4 Adult 2415
```

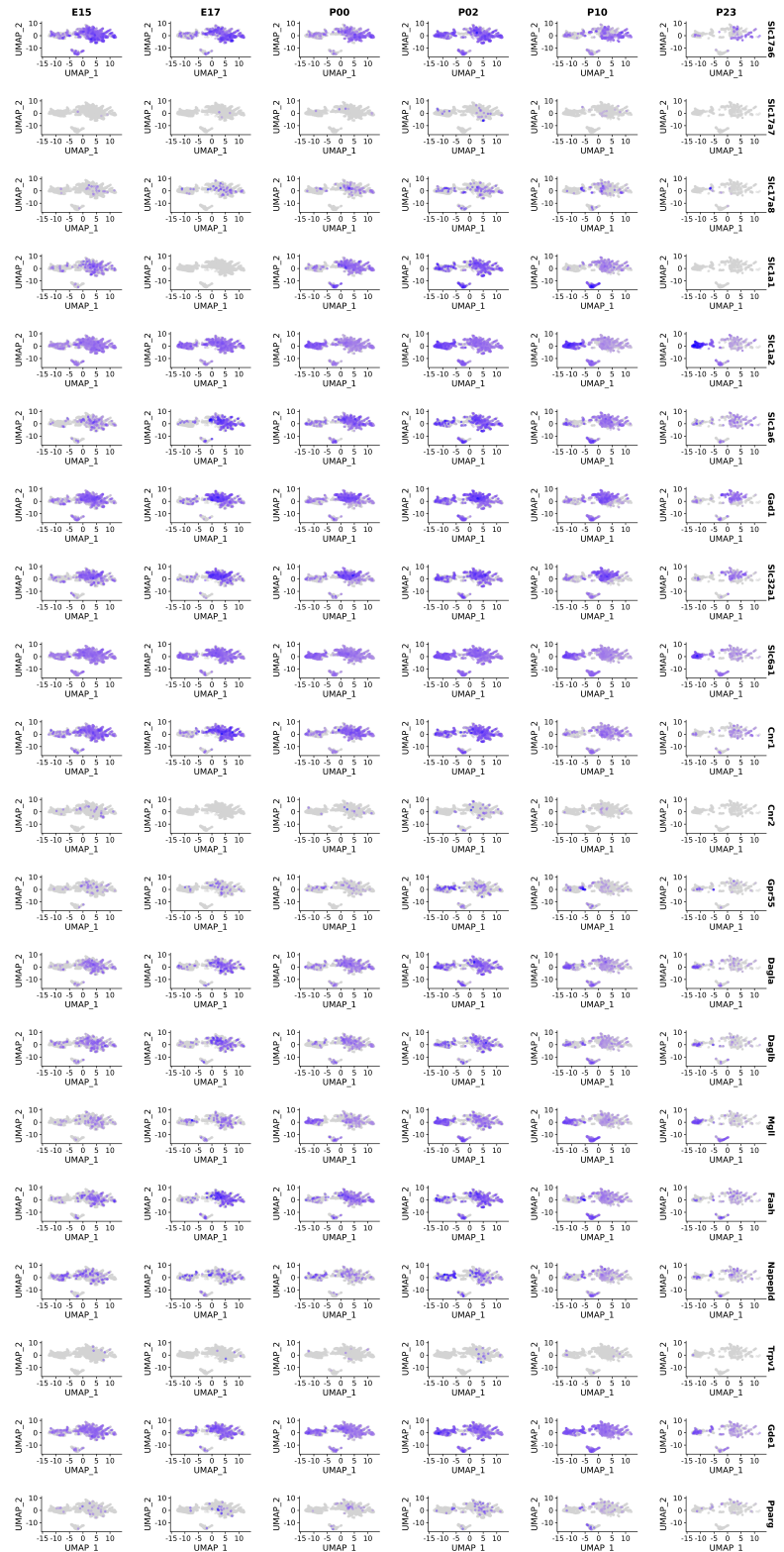
```
rar2020.srt.pub$age <-
  plyr::mapvalues(
    x = rar2020.srt.pub$age,
    from = c("E15", "E17", "P0", "P2", "3P2", "1P10", "P10", "P23"),
    to = c("E15", "E17", "P00", "P02", "P02", "P10", "P10", "P23")
  )
```

```
rar2020.srt.pub$age %>% forcats::fct_count()
```

```
# A tibble: 6 x 2
```

```
  f      n  
  <fct> <int>  
1 E15    8290  
2 E17   11213  
3 P00    7492  
4 P02   12824  
5 P10    8965  
6 P23    2415
```

```
FeaturePlot(  
  rar2020.srt.pub,  
  features = c(neurotrans, cnbn),  
  label = F,  
  blend = F,  
  order = TRUE,  
  pt.size = 1.2,  
  raster.dpi = c(1024, 1024),  
  alpha = 0.5,  
  split.by = "age"  
)
```



```

sbs_mtx <-
  rar2020.srt.pub@assays$RNA@data %>%
  as.data.frame() %>%
  t()
rownames(sbs_mtx) <- colnames(rar2020.srt.pub)

# Filter features
filt_low_genes <-
  colSums(sbs_mtx) %>%
  .[, > quantile(., 0.4)] %>%
  names()
sbs_mtx %<>% .[, filt_low_genes]

min_filt_vector2 <-
  sbs_mtx %>%
  as_tibble() %>%
  select(all_of(filt_low_genes)) %>%
  summarise(across(.fns = ~ quantile(., .005))) %>%
  as.list() %>%
  map(as.double) %>%
  simplify() %>%
  .[filt_low_genes]

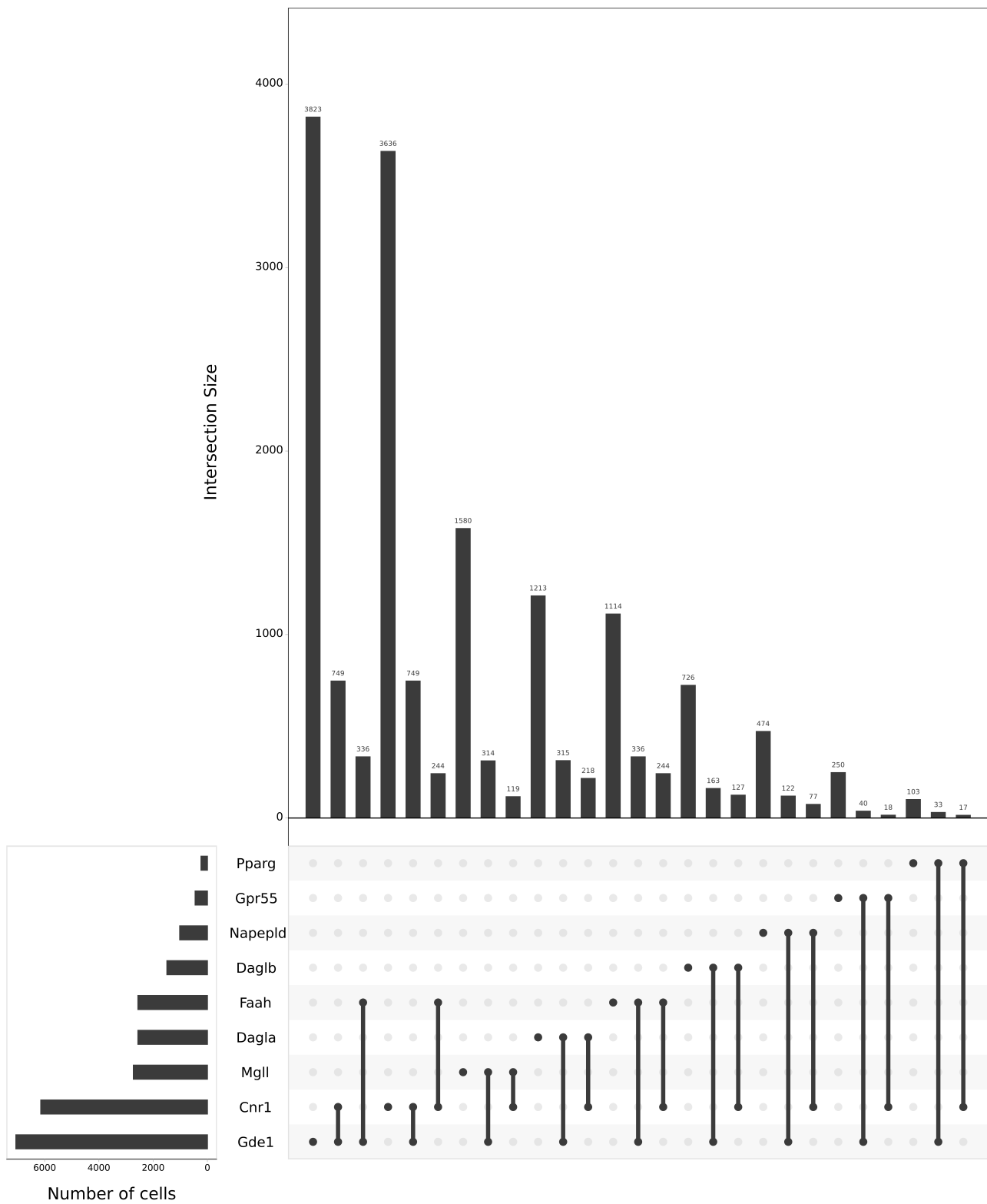
# Prepare table of intersection sets analysis
content_sbs_mtx <-
  (sbs_mtx > min_filt_vector2) %>%
  as_tibble() %>%
  mutate_all(as.numeric)

upset(
  as.data.frame(content_sbs_mtx),
  order.by = "freq",
  group.by = "sets",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,
  point.size = 3.5, line.size = 2,
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
  nsets = 30,

```



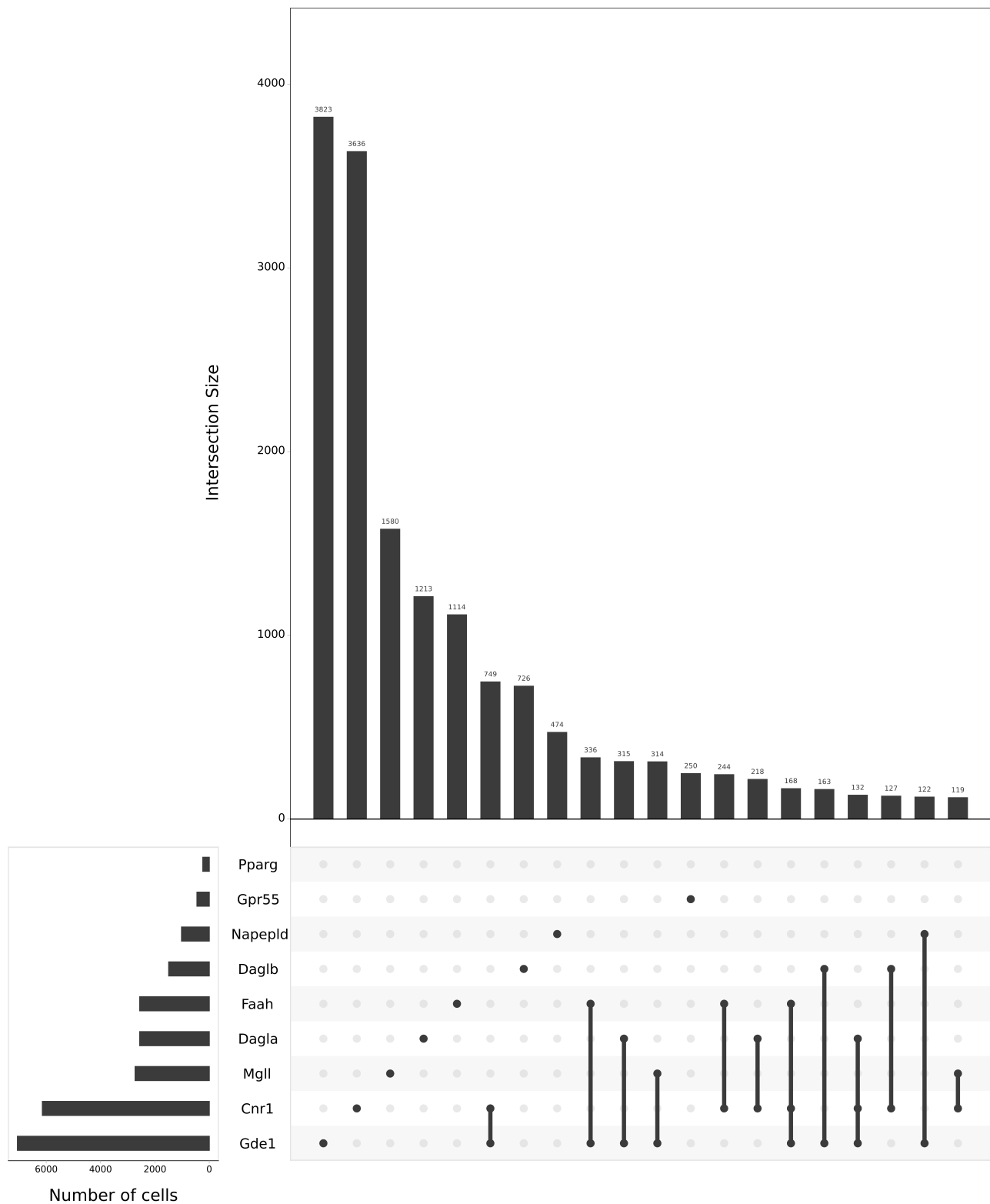
```
nintersects = 30,  
sets = c(cnbn) %>%  
  [. %in% colnames(content_sbs_mtx)],  
empty.intersections = NULL  
)
```



```

upset(
  as.data.frame(content_sbs_mtx),
  order.by = "freq",
  sets.x.label = "Number of cells",
  number.angles = 0,
  point.size = 3.5, line.size = 2,
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
  nsets = 15,
  sets = c(cnbn) %>%
    [. %in% colnames(content_sbs_mtx)],
  nintersects = 20,
  empty.intersections = NULL
)

```



```
sbs_mtx_full <- content_sbs_mtx |>
  select(any_of(c(neurotrans, cnbn))) |>
  dplyr::bind_cols(rar2020.srt.pub@meta.data)
```

```
sbs_mtx_full |> glimpse()
```

Rows: 51,199

Columns: 38

```
$ Slc17a6      <dbl> 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ Slc17a8      <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ Slc1a1       <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ Slc1a2       <dbl> 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ Slc1a6       <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ Gad1         <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, ~
$ Slc32a1      <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ Slc6a1       <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, ~
$ Cnr1         <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ Gpr55        <dbl> 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ Dagla        <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ Daglb        <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ Mgll         <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ Faah         <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ Napepld      <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ Gde1         <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, ~
$ Pparg        <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
$ nGene        <int> 1652, 782, 447, 1706, 1106, 894, 727, 734, 669, 617, ~
$ nUMI         <dbl> 2787, 1090, 544, 2709, 1817, 1220, 995, 1036, 920, 86~
$ orig.ident   <fct> Hypothalamus, Hypothalamus, Hypothalamus, Hypothalamu~
$ res.0.2      <chr> "23", "23", "23", "23", "23", "23", "23", "23", "23", ~
$ res.0.4      <chr> "34", "34", "34", "34", "34", "34", "34", "34", "34", ~
$ res.0.8      <chr> "42", "42", "42", "42", "42", "42", "42", "42", "42", ~
$ res.1.2      <chr> "47", "47", "47", "47", "47", "47", "47", "47", "47", ~
$ res.2        <chr> "54", "54", "54", "54", "54", "54", "54", "54", "54", ~
$ tree.ident   <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, ~
$ pro_Inter    <chr> "41", "41", "41", "41", "41", "41", "41", "41", "41", ~
$ pro_Enter    <chr> "41", "41", "41", "41", "41", "41", "41", "41", "41", ~
$ tree_final   <fct> 19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 19, 1~
$ subtree      <fct> 41, 41, 41, 41, 41, 41, 41, 41, 41, 41, 41, 41, 41, 4~
$ prim_walktrap <fct> 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 3~
$ umi_per_gene <dbl> 1.687046, 1.393862, 1.217002, 1.587925, 1.642857, 1.3~
```

```

$ log_umi_per_gene <dbl> 0.22712693, 0.14421974, 0.08529138, 0.20082998, 0.215~
$ nCount_RNA      <dbl> 2787, 1090, 544, 2709, 1817, 1220, 995, 1036, 920, 86~
$ nFeature_RNA    <int> 1652, 782, 447, 1706, 1106, 894, 727, 734, 669, 617, ~
$ wtrees          <fct> 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 38, 3~
$ age             <chr> "P23", "P02", "P02", "P02", "P02", "P02", "P02", "P02~
$ stage           <ord> Adult, Neonatal, Neonatal, Neonatal, Neonatal, Neonat~

```

Prepare query mapping between datasets

```

srt <- NormalizeData(srt)
srt <- FindVariableFeatures(srt, selection.method = "vst", nfeatures = 3000)
# all.genes <- rownames(srt)
# srt <- ScaleData(srt, features = all.genes)
srt <- ScaleData(srt)

```

```

hypoth.anchors <- FindTransferAnchors(
  reference = rar2020.srt.pub, query = srt, dims = 1:30,
  reference.reduction = "pca"
)
predictions <- TransferData(anchorset = hypoth.anchors, refdata = rar2020.srt.pub$wtrees)
srt <- AddMetaData(srt, metadata = predictions)
table(srt$predicted.id)

```

1	10	11	12	13	15	16	17	19	2	20
3700	1	2198	225	545	58	218	19	100308	9886	924
22	23	26	27	28	29	3	38	4	5	6
163	11	115	24	76	481	951	49	156	2895	4756
7	8	9								
199	47	1								

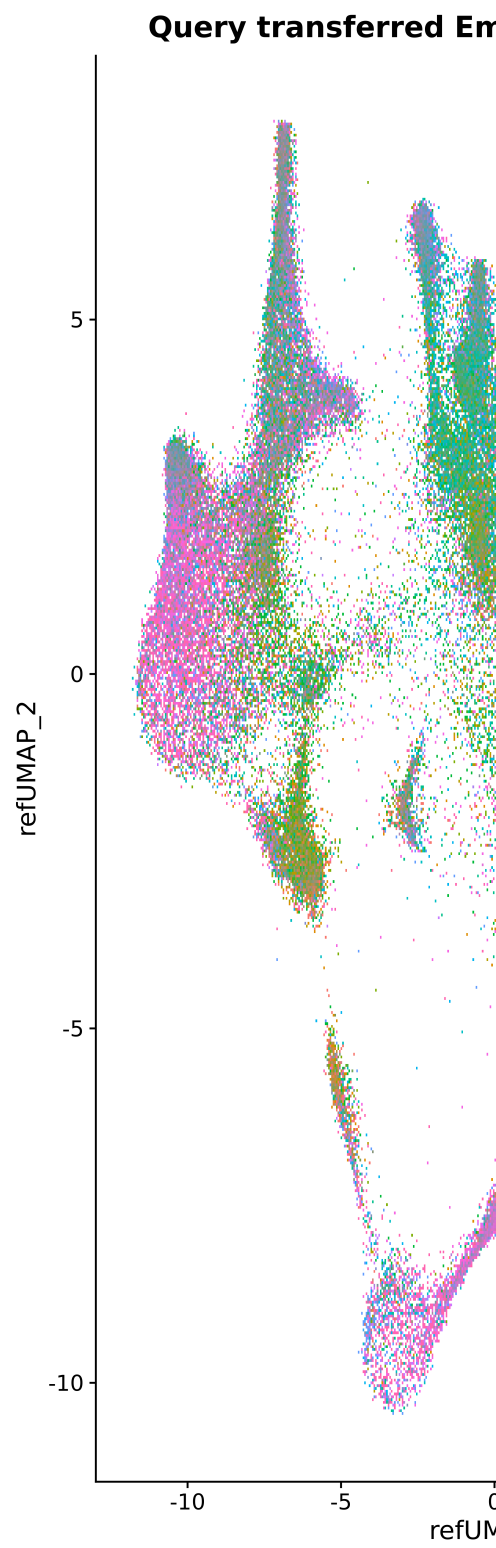
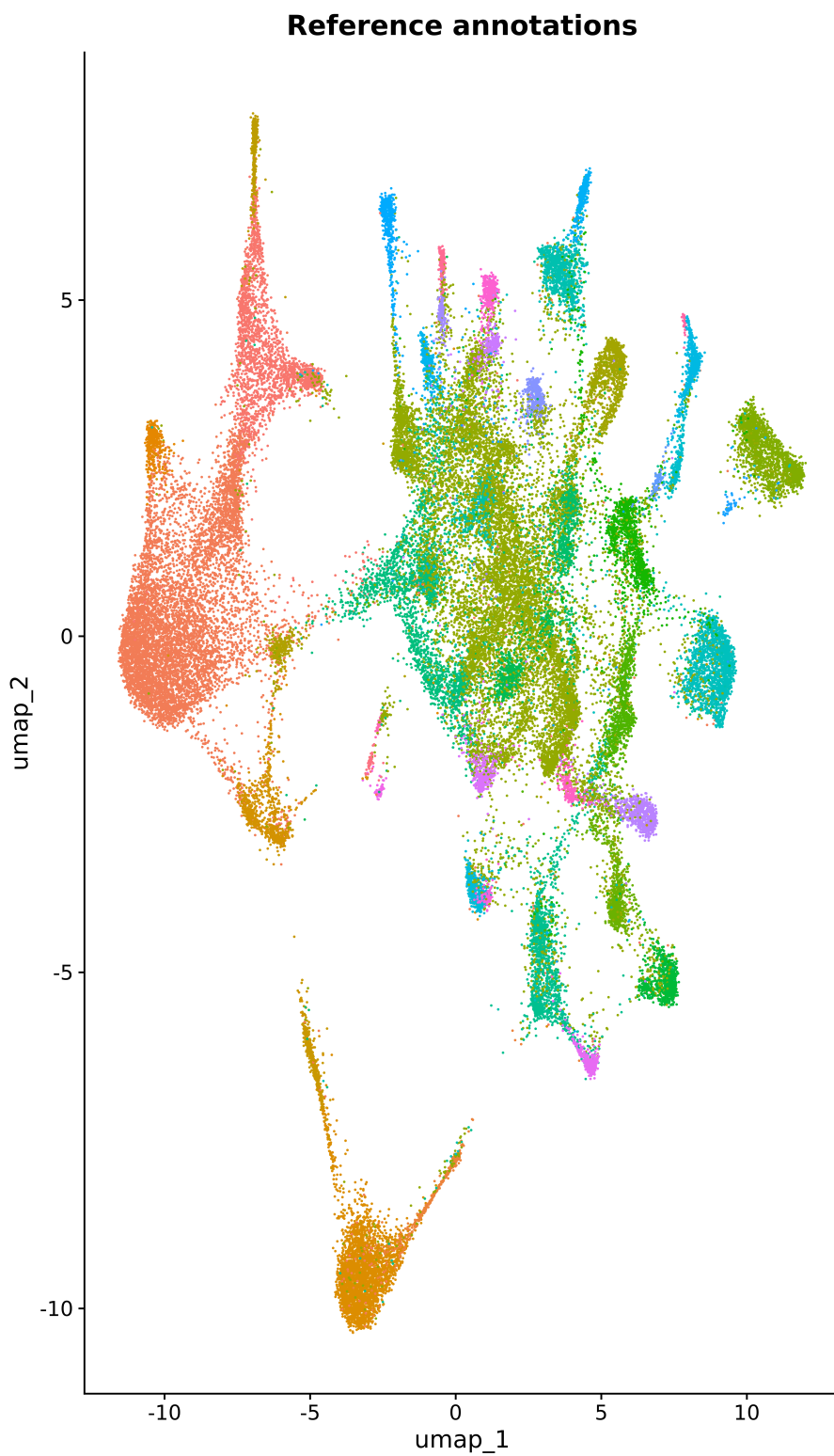
```

rar2020.srt.pub <- RunUMAP(rar2020.srt.pub, dims = 1:30, reduction = "pca", return.mds = FALSE)
srt <- IntegrateEmbeddings(
  anchorset = hypoth.anchors, reference = rar2020.srt.pub, query = srt,
  new.reduction.name = "ref.pca"
)
srt <- ProjectUMAP(
  query = srt, query.reduction = "ref.pca", reference = rar2020.srt.pub,

```

```
    reference.reduction = "pca", reduction.model = "umap"
  )
  Idents(srt) <- srt$Cluster

p1 <- DimPlot(rar2020.srt.pub,
  reduction = "umap", group.by = "wtree", label = F
) + NoLegend() + ggtitle("Reference annotations")
p2 <- DimPlot(srt,
  reduction = "ref.umap", group.by = "Age", label = F
) + NoLegend() + ggtitle("Query transferred Embedding (more ages)")
p1 + p2
```

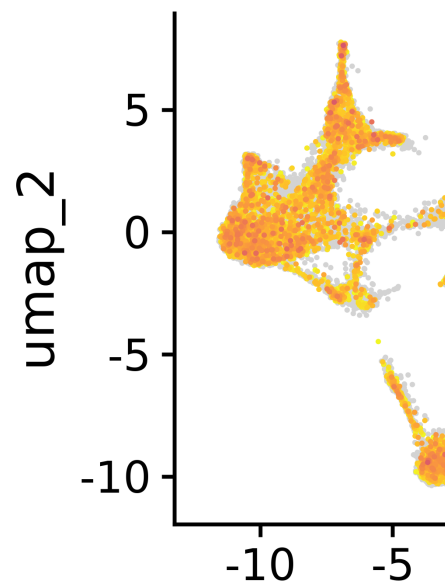
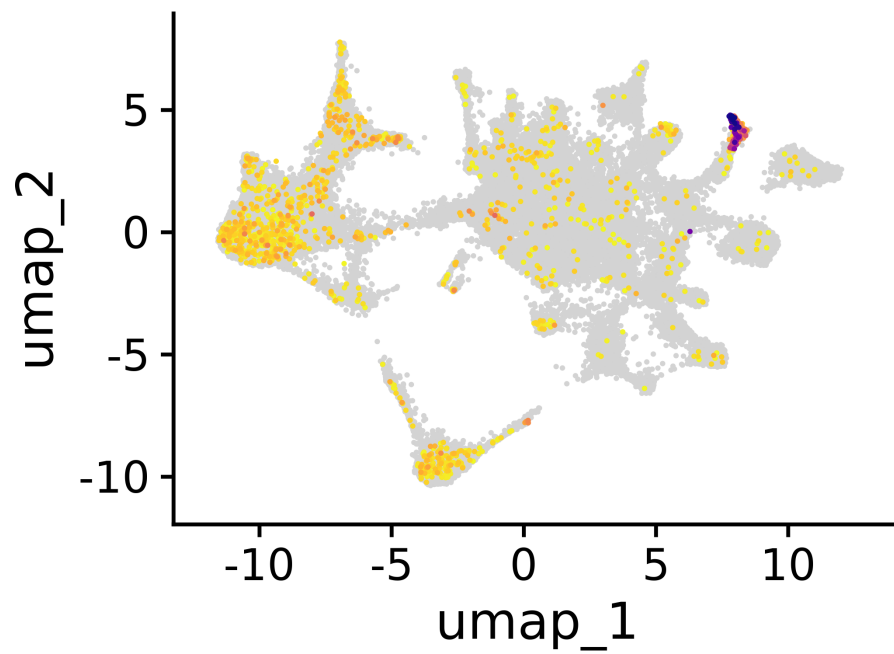



```

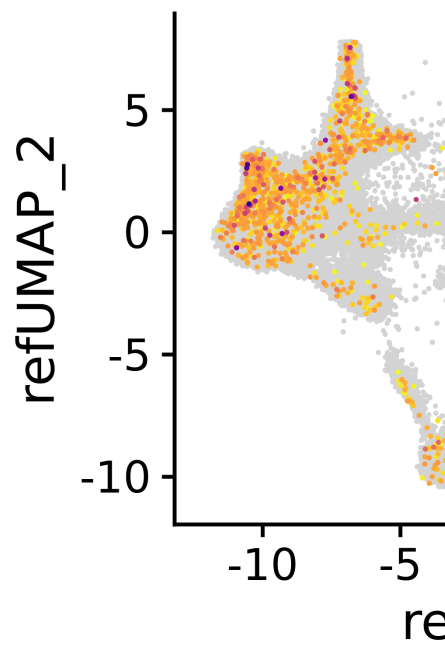
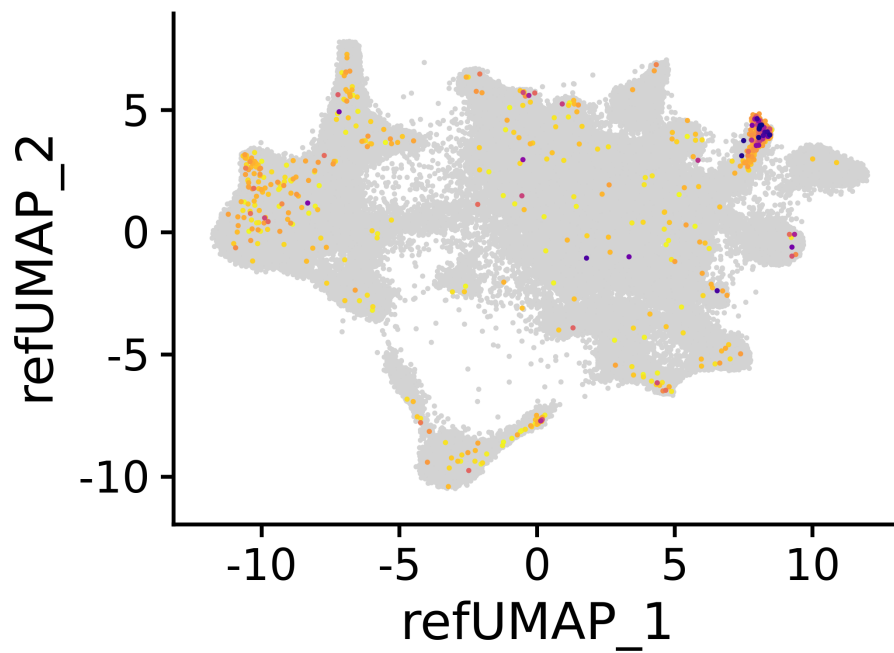
p1 <- FeaturePlot_scCustom(
  rar2020.srt.pub,
  reduction = "umap",
  features = c(
    "Oxt",
    "Avp",
    "Sst",
    "Crh",
    "Trh"),
  label = F,
  num_columns = 5,
  min.cutoff = 'q05',
  na_cutoff = 2
) * NoLegend()
p2 <- FeaturePlot_scCustom(
  srt,
  reduction = "ref.umap",
  features = c(
    "Oxt",
    "Avp",
    "Sst",
    "Crh",
    "Trh"),
  label = FALSE,
  num_columns = 5,
  min.cutoff = 'q05',
  na_cutoff = 2
) * NoLegend()
(p1 / p2)

```

Oxt



Oxt

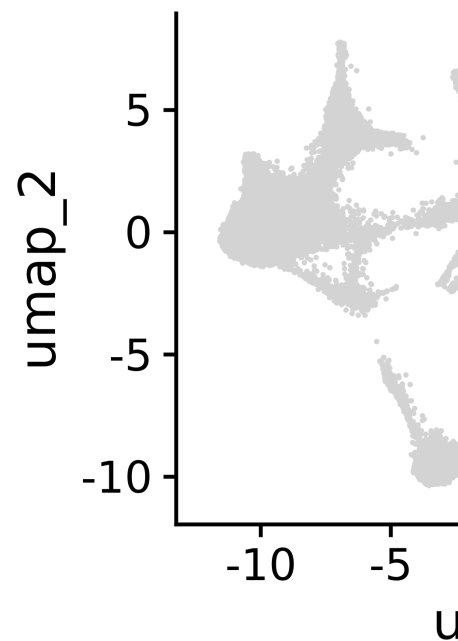
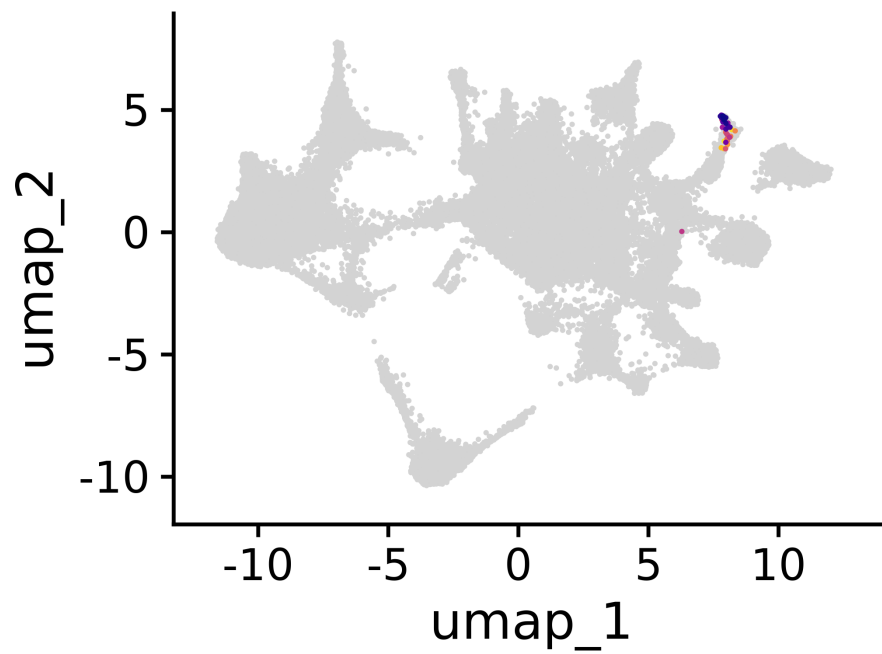


```

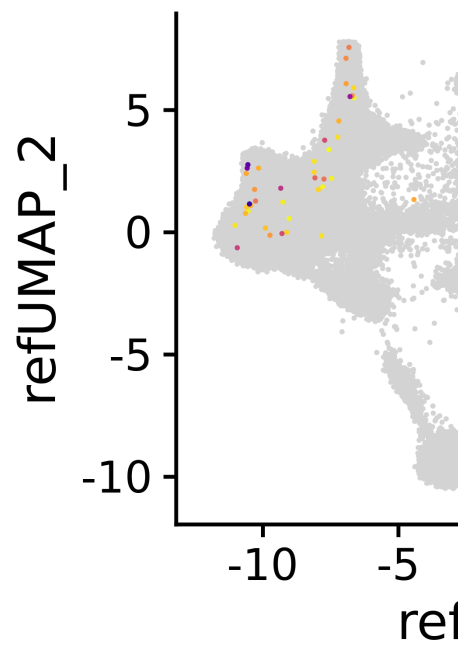
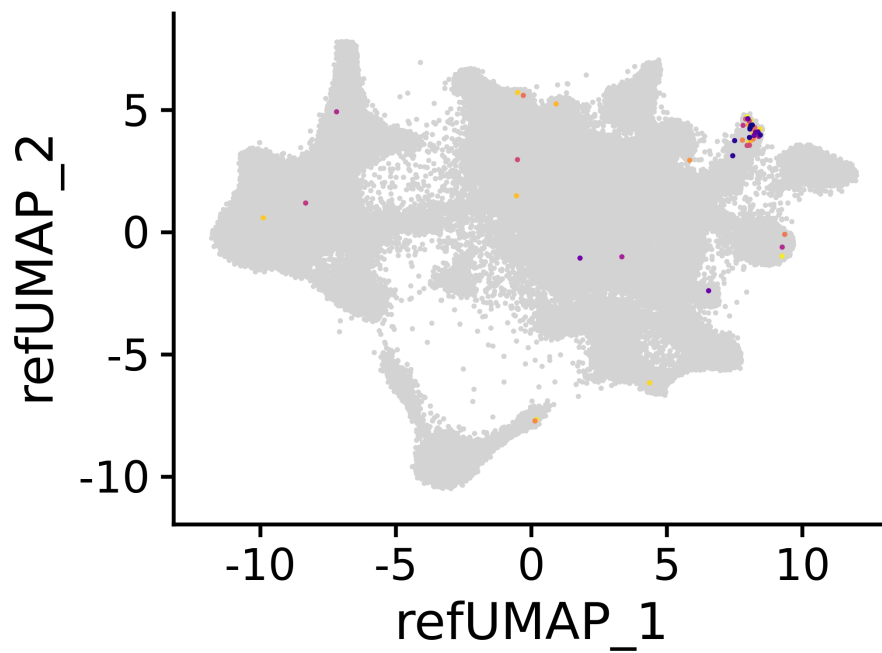
p1 <- FeaturePlot_scCustom(
  rar2020.srt.pub,
  reduction = "umap",
  features = c(
    "Oxt",
    "Avp"),
  label = F,
  num_columns = 2,
  min.cutoff = 'q05',
  na_cutoff = 5
) * NoLegend()
p2 <- FeaturePlot_scCustom(
  srt,
  reduction = "ref.umap",
  features = c(
    "Oxt",
    "Avp"),
  label = FALSE,
  num_columns = 2,
  min.cutoff = 'q05',
  na_cutoff = 5
) * NoLegend()
(p1 / p2)

```

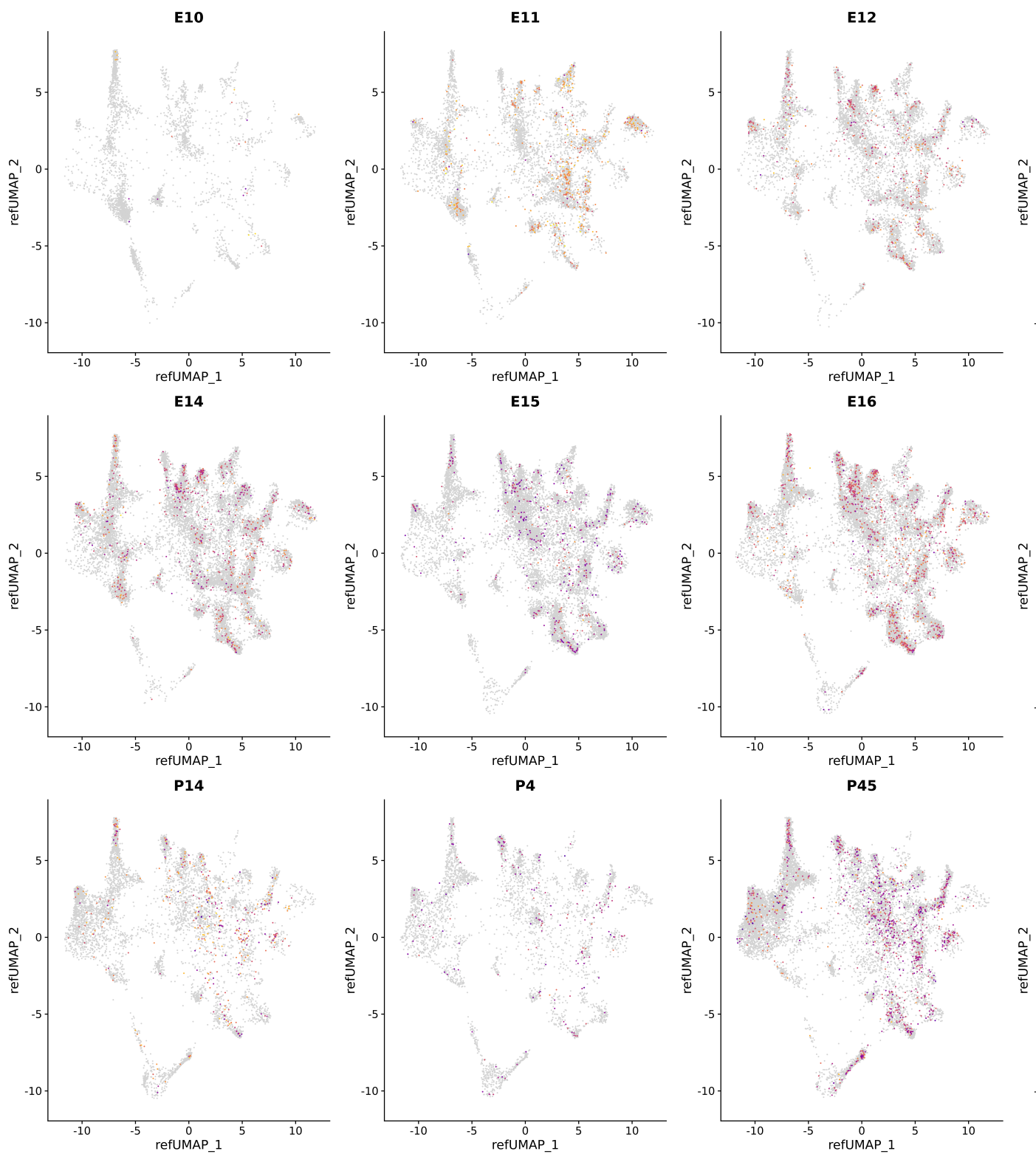
Oxt



Oxt



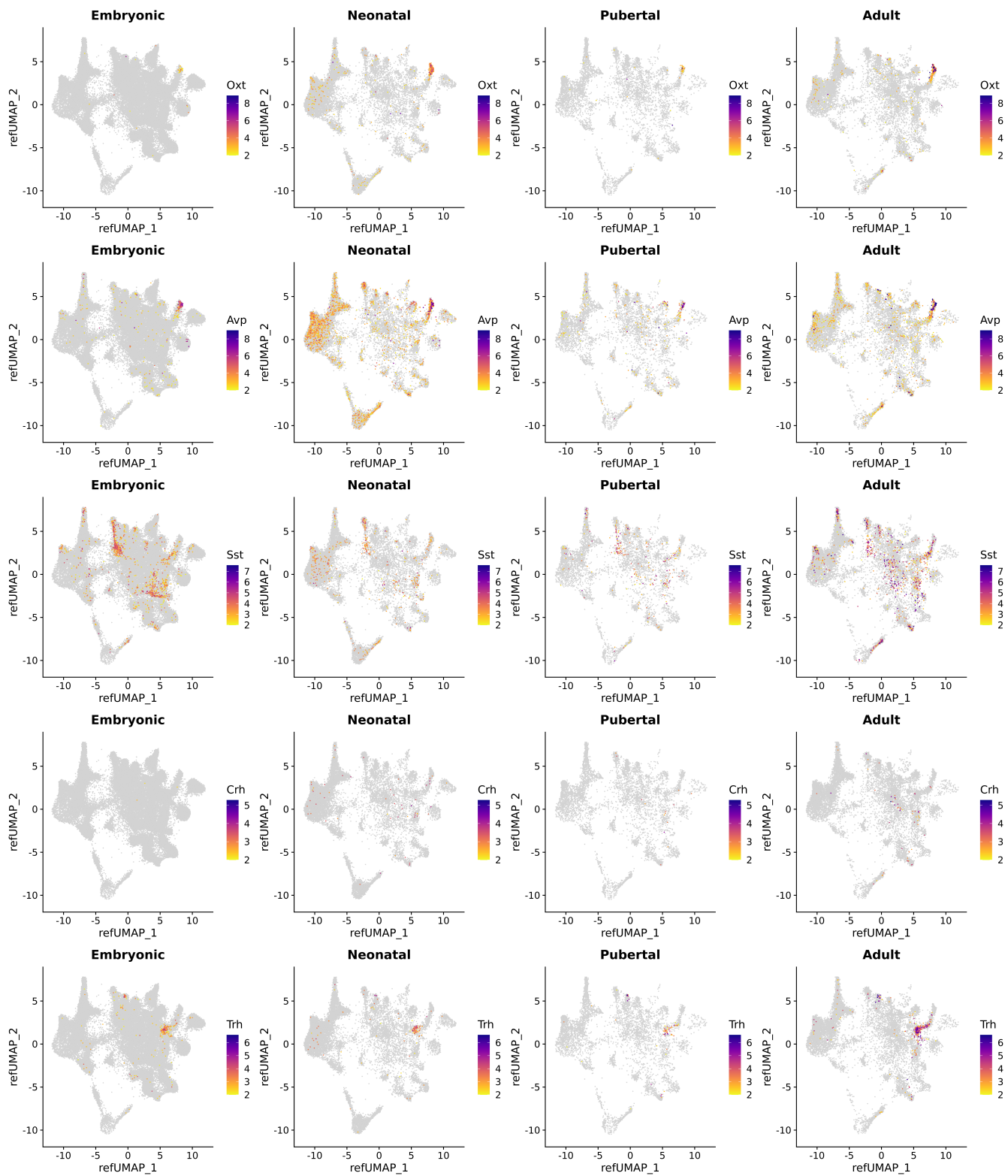
```
FeaturePlot_scCustom(srt, reduction = "ref.umap", features = c("Cnr1"), split.by = "
```



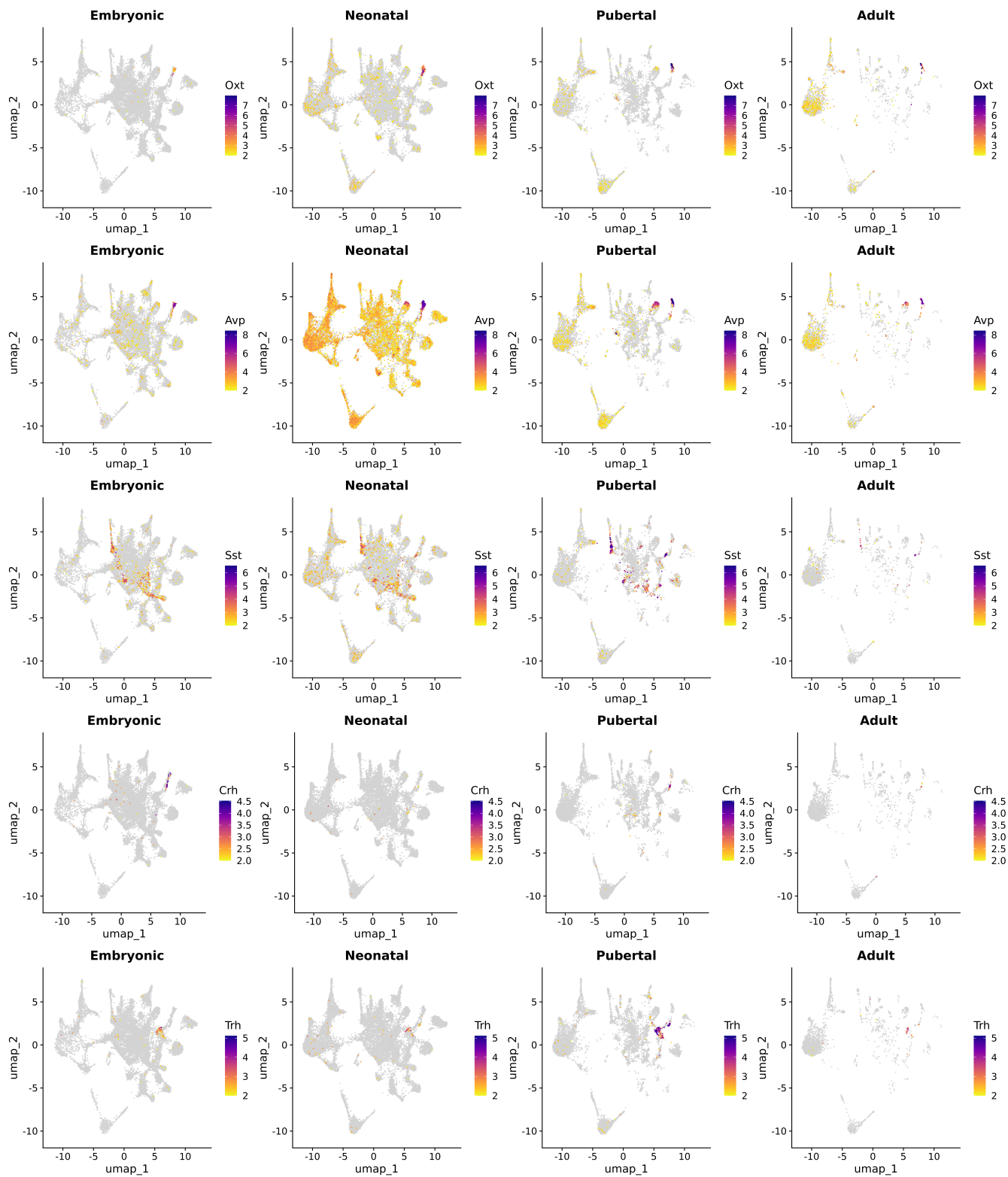
```

srt$stage <-
srt$Age %>%
forcats::fct_collapse(
  Embryonic = c(
    "E10", "E11", "E12", "E13",
    "E14", "E15", "E16", "E18"
  ),
  Neonatal = c("P4", "P8"),
  Pubertal = c("P14"),
  Adult = c("P45")
)
srt$stage %<>% factor(levels = c("Embryonic", "Neonatal", "Pubertal", "Adult"), orde
FeaturePlot_scCustom(
  srt,
  reduction = "ref.umap",
  features = c(
    "Oxt",
    "Avp",
    "Sst",
    "Crh",
    "Trh"),
  split.by = "stage",
  min.cutoff = 'q05',
  na_cutoff = 2,
  label = F,
  num_columns = 4
) * NoLegend()

```




```
FeaturePlot_scCustom(  
  rar2020.srt.pub,  
  reduction = "umap",  
  features = c(  
    "Oxt",  
    "Avp",  
    "Sst",  
    "Crh",  
    "Trh"),  
  split.by = "stage",  
  min.cutoff = 'q05',  
  na_cutoff = 2,  
  label = F,  
  num_columns = 4  
) * NoLegend()
```



```

if (!file.exists(here(data_dir, "kim2020_pvn_neurons.txt"))) {
  plot <- DimPlot(object = srt, reduction = "ref.umap")
  srt <- CellSelector(plot = plot, object = srt, ident = "SelectedCells")

  selected_cells <- Cells(subset(srt, ident = "SelectedCells"))
  write_lines(selected_cells, file = here(data_dir, "kim2020_pvn_neurons.txt"))
}
selected_cells <- read_lines(here(data_dir, "kim2020_pvn_neurons.txt"))
srt <- subset(srt, cells = selected_cells)
srt <- subset(srt, subset = refUMAP_1 > 4 & refUMAP_2 > -1)

srt@meta.data <- srt@meta.data |> rename(wtree = predicted.id, age = Age)

srt

```

An object of class Seurat
 27998 features across 4555 samples within 1 assay
 Active assay: RNA (27998 features, 3000 variable features)
 3 layers present: counts, data, scale.data
 3 dimensional reductions calculated: umap, ref.pca, ref.umap

Intersection sets analysis

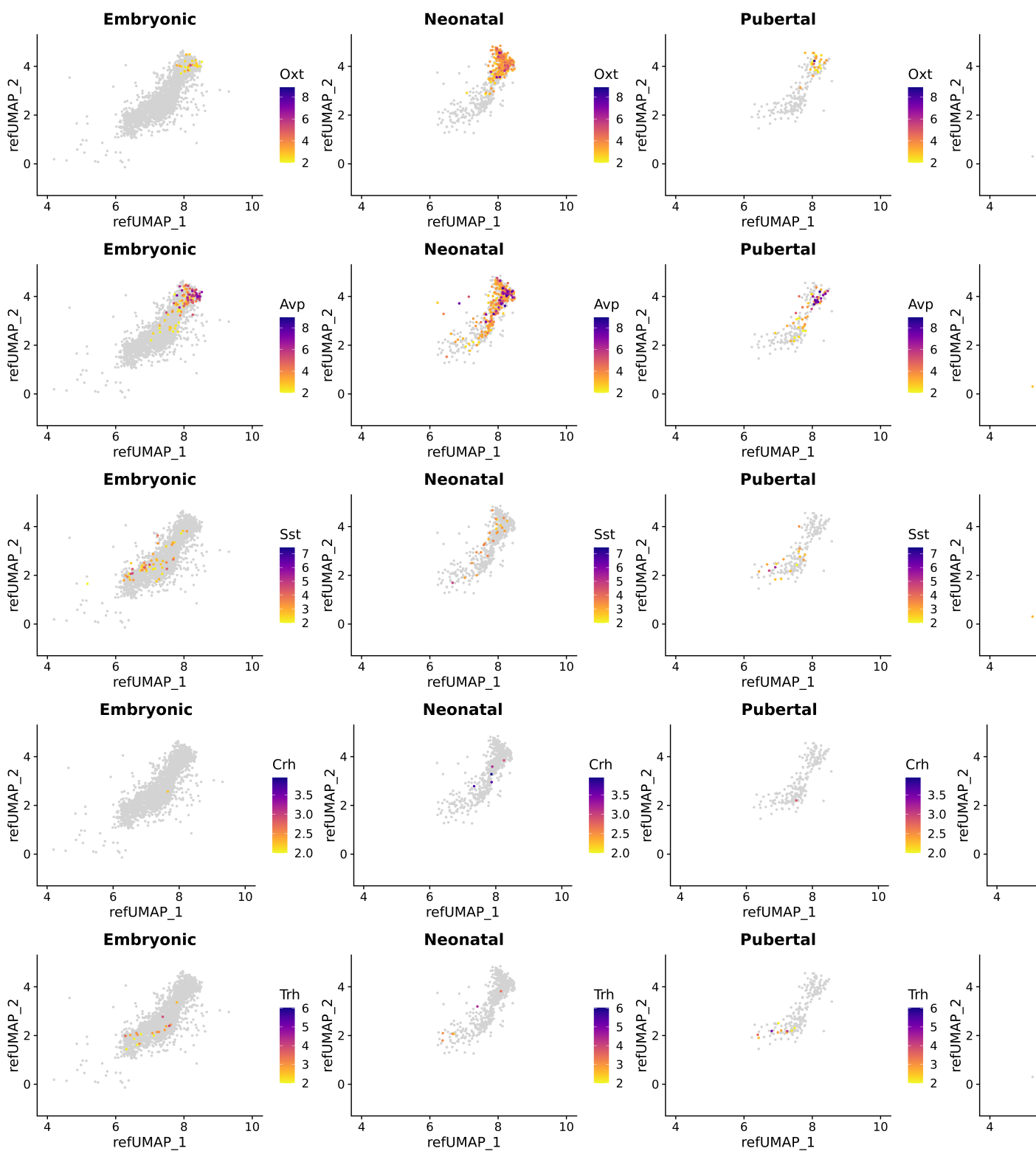
PVN Neurons from Kim et al. 2020, Nature Communications

```

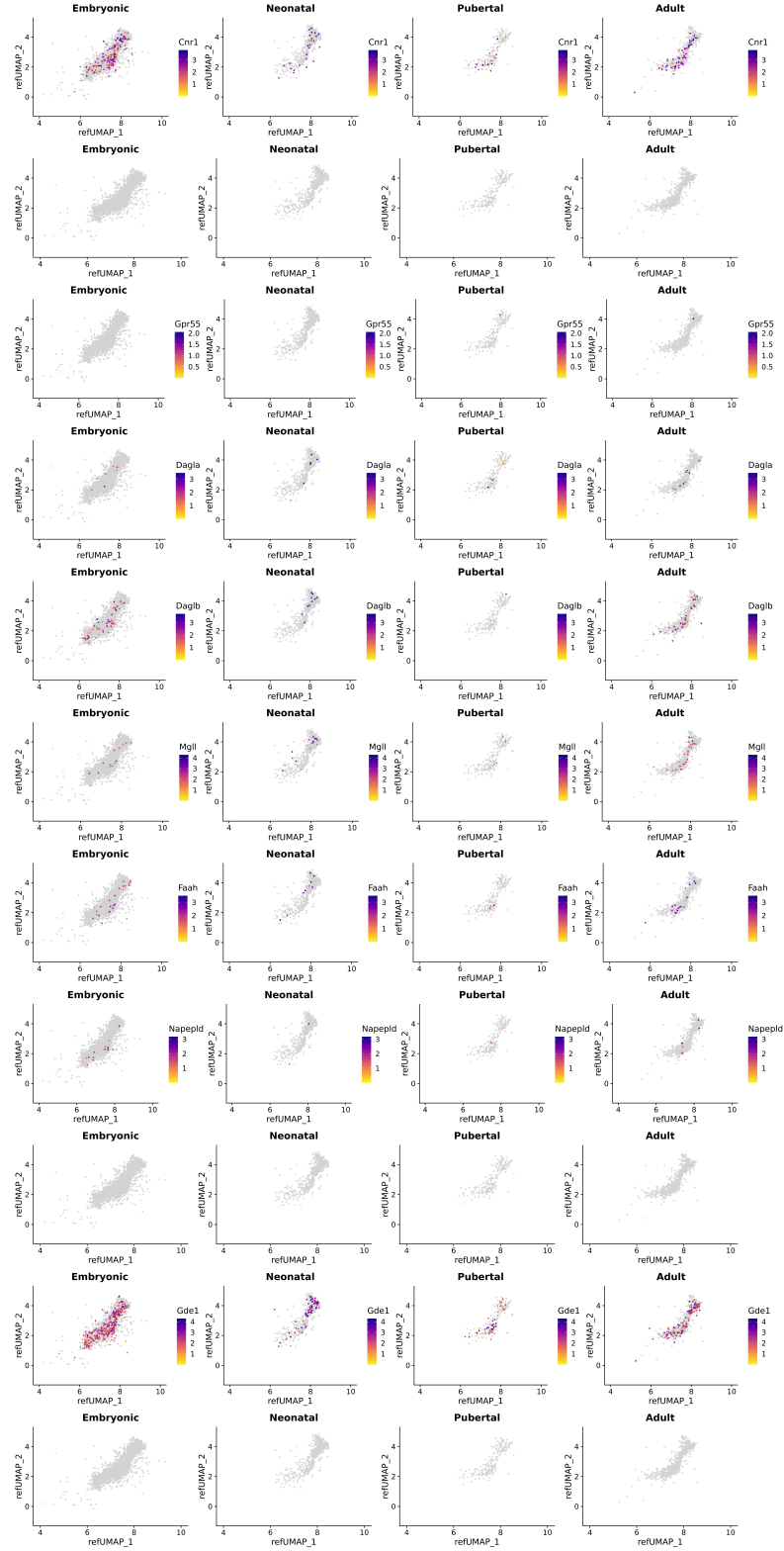
FeaturePlot_scCustom(
  srt,
  reduction = "ref.umap",
  features = c(
    "Oxt",
    "Avp",
    "Sst",
    "Crh",
    "Trh"),
  split.by = "stage",
  na_cutoff = 2,
  label = F,
  num_columns = 4
)

```

```
) * NoLegend()
```



```
FeaturePlot_scCustom(srt, reduction = "ref.umap", features = c(cnb), split.by = "st
```



```

sbs_mtx <-
  srt@assays$RNA@layers$data %>%
  as.data.frame() %>%
  t()

rownames(sbs_mtx) <- colnames(srt)
colnames(sbs_mtx) <- rownames(srt)

# Filter features
filt_low_genes <-
  colSums(sbs_mtx) %>%
  .[, > quantile(., 0.4)] %>%
  names()
sbs_mtx %<>% .[, filt_low_genes]

min_filt_vector2 <-
  sbs_mtx %>%
  as_tibble() %>%
  select(all_of(filt_low_genes)) %>%
  summarise(across(.fns = ~ quantile(.x, .005))) %>%
  as.list() %>%
  map(as.double) %>%
  simplify() %>%
  .[filt_low_genes]

# Prepare table of intersection sets analysis
content_sbs_mtx_kim <-
  (sbs_mtx > min_filt_vector2) %>%
  as_tibble() %>%
  mutate_all(as.numeric) %>%
  bind_cols(
    srt@meta.data |> select(wtree, age, stage)
  )

```

All

```

upset(
  as.data.frame(content_sbs_mtx_kim),
  order.by = "freq",

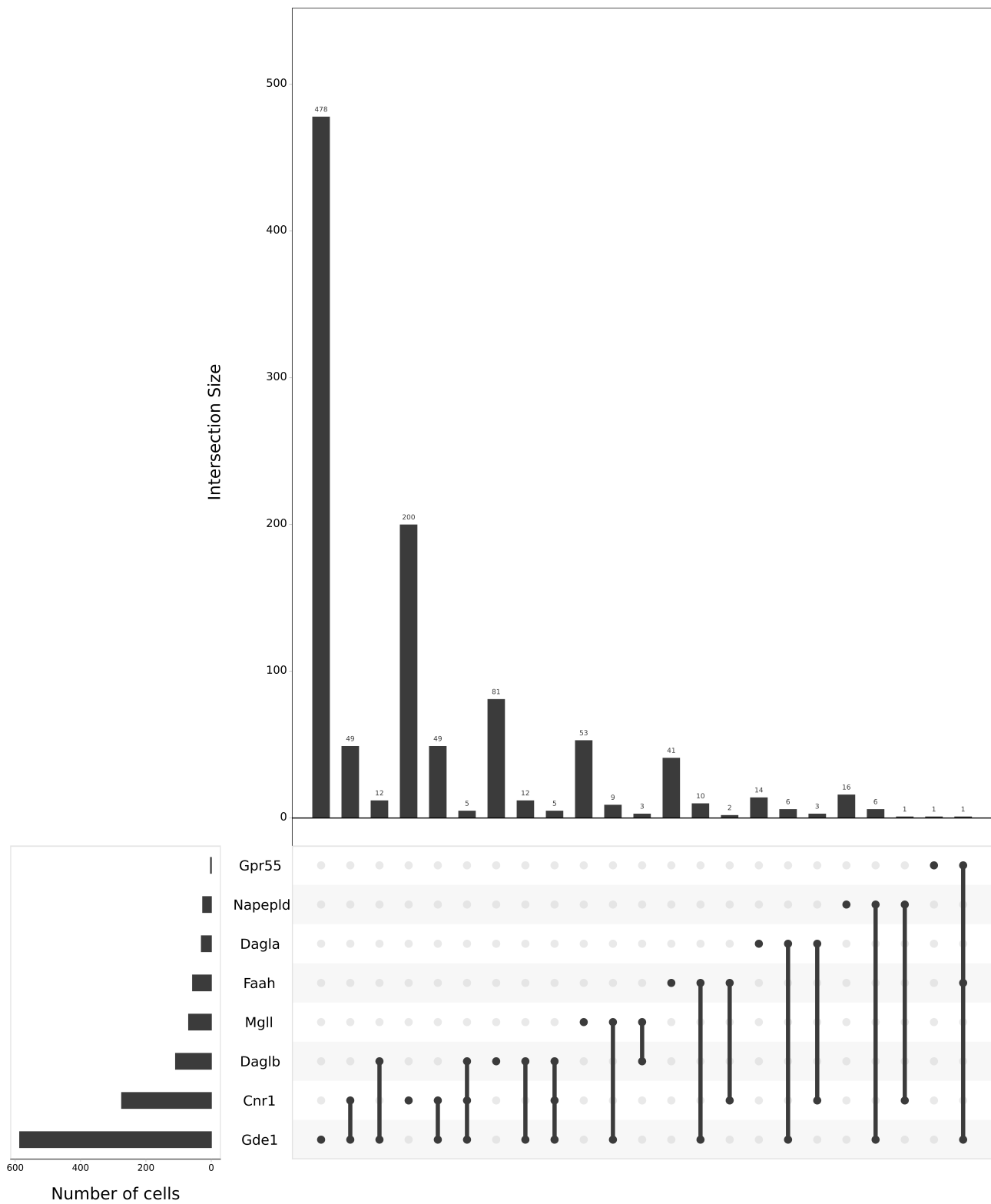
```



```

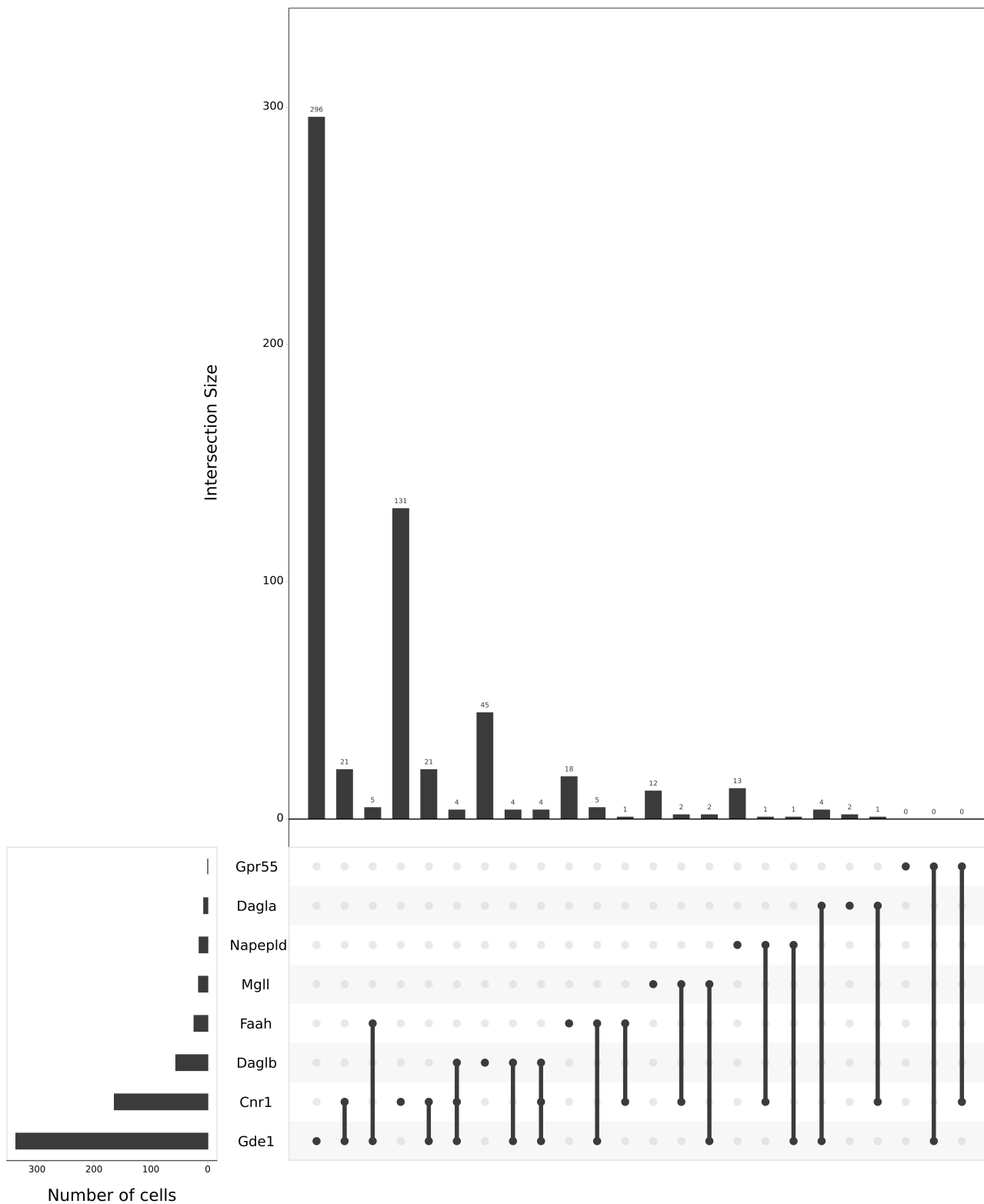
group.by = "sets",
cutoff = 3,
sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c(cnbn) %>%
  [. %in% colnames(content_sbs_mtx_kim)],
empty.intersections = NULL
)

```



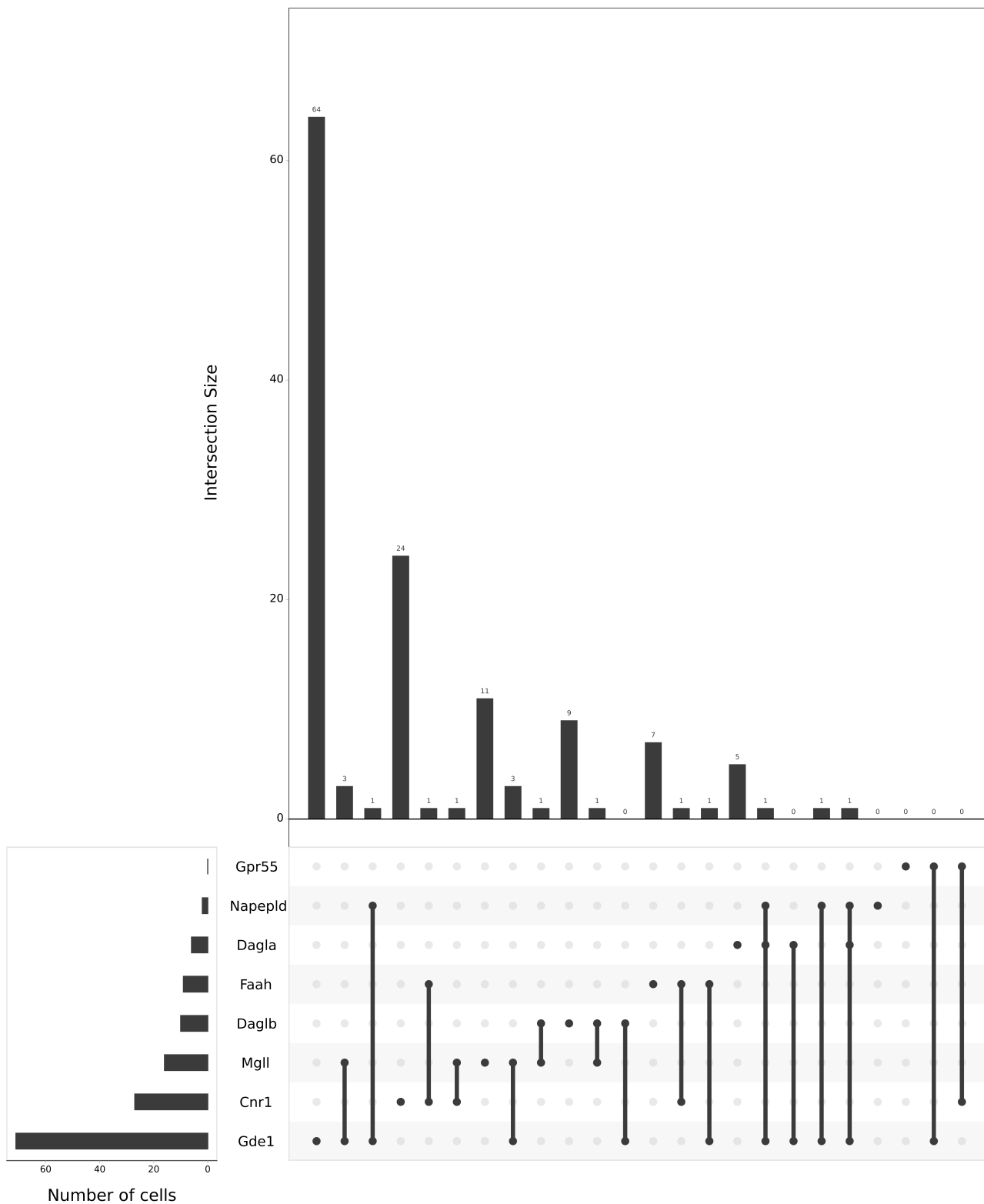
Embryonic

```
upset(
  as.data.frame(
    content_sbs_mtx_kim |>
      filter(stage == "Embryonic") |>
        select(
          c(cnbn) %>% .[. %in% colnames(content_sbs_mtx_kim)]
        )
  ),
  order.by = "freq",
  group.by = "sets",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,
  point.size = 3.5, line.size = 2,
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
  sets = c(cnbn) %>%
    .[. %in% colnames(content_sbs_mtx_kim)],
  empty.intersections = F
)
```



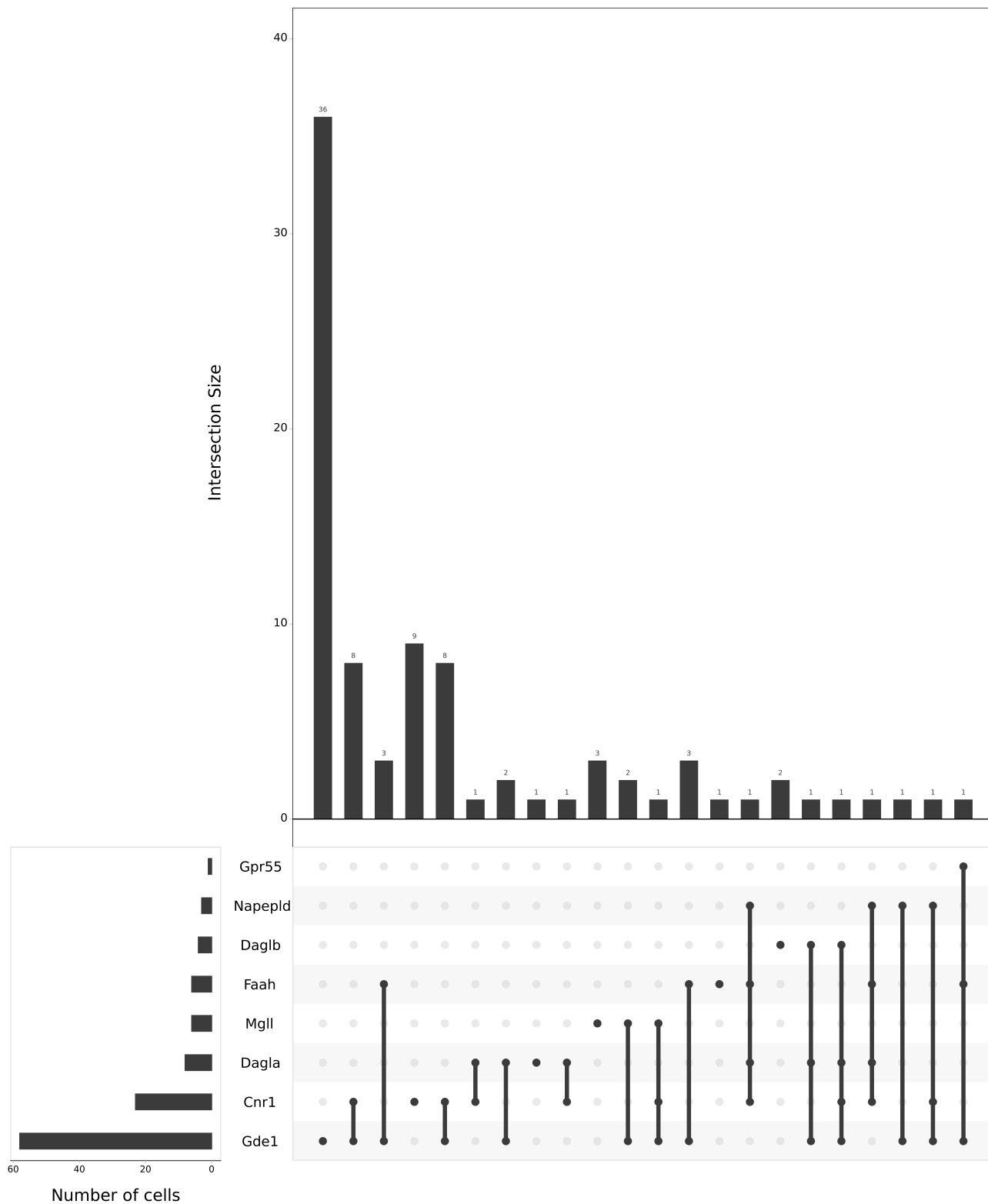
Neonatal

```
upset(
  as.data.frame(
    content_sbs_mtx_kim |>
      filter(stage == "Neonatal") |>
        select(
          c(cnbn) %>% [. %in% colnames(content_sbs_mtx_kim)]
        )
  ),
  order.by = "freq",
  group.by = "sets",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,
  point.size = 3.5, line.size = 2,
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
  nsets = 30,
  nintersects = 30,
  sets = c(cnbn) %>%
    [. %in% colnames(content_sbs_mtx_kim)],
  empty.intersections = F
)
```



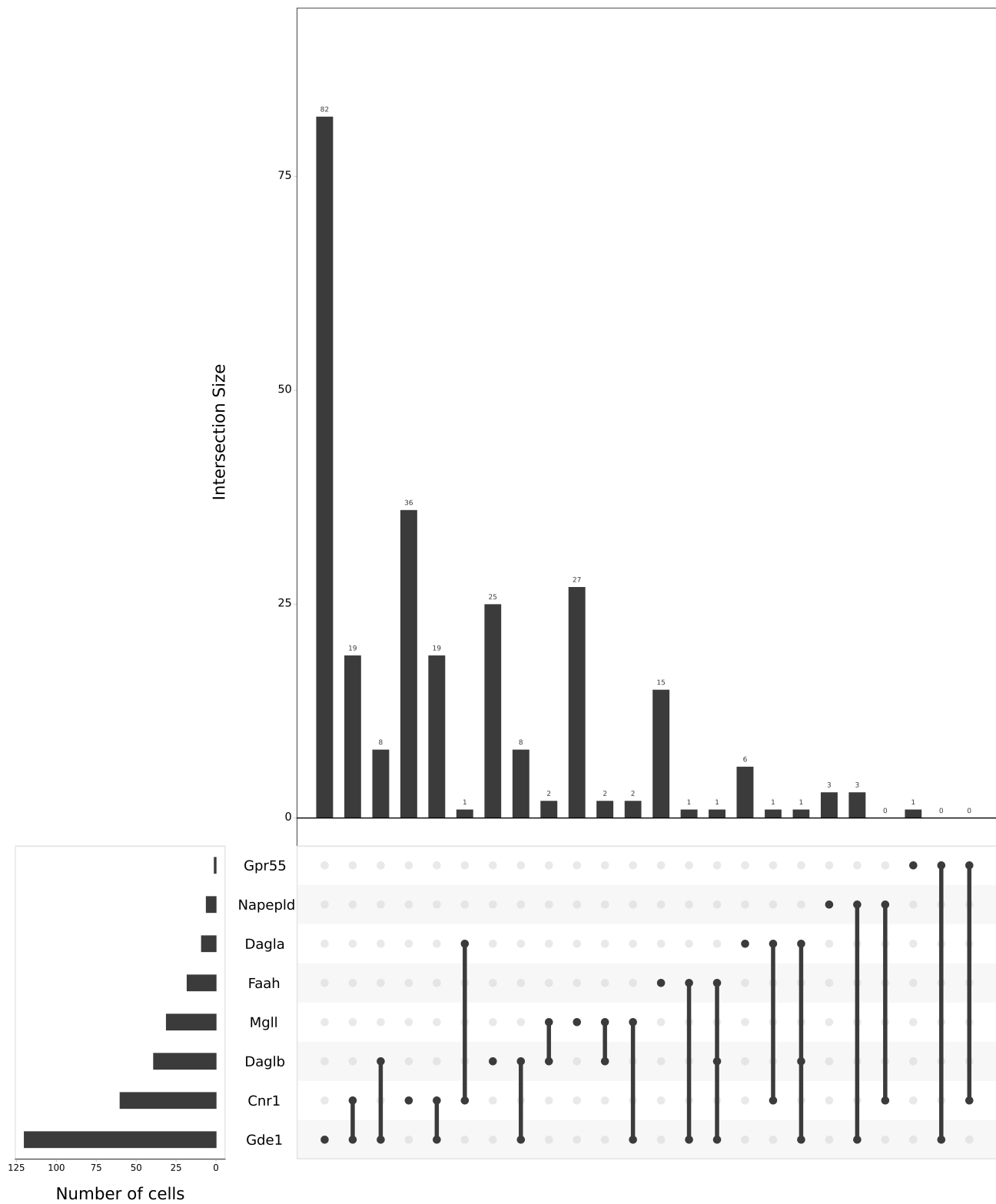
Pubertal

```
upset(  
  as.data.frame(  
    content_sbs_mtx_kim |>  
    filter(stage == "Pubertal") |>  
    select(  
      c(cnbn) %>% .[. %in% colnames(content_sbs_mtx_kim)]  
    )  
  ),  
  order.by = "freq",  
  group.by = "sets",  
  cutoff = 3,  
  sets.x.label = "Number of cells",  
  number.angles = 0,  
  point.size = 3.5, line.size = 2,  
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),  
  nsets = 30,  
  nintersects = 30,  
  sets = c(cnbn) %>%  
    .[. %in% colnames(content_sbs_mtx_kim)],  
  empty.intersections = NULL  
)
```



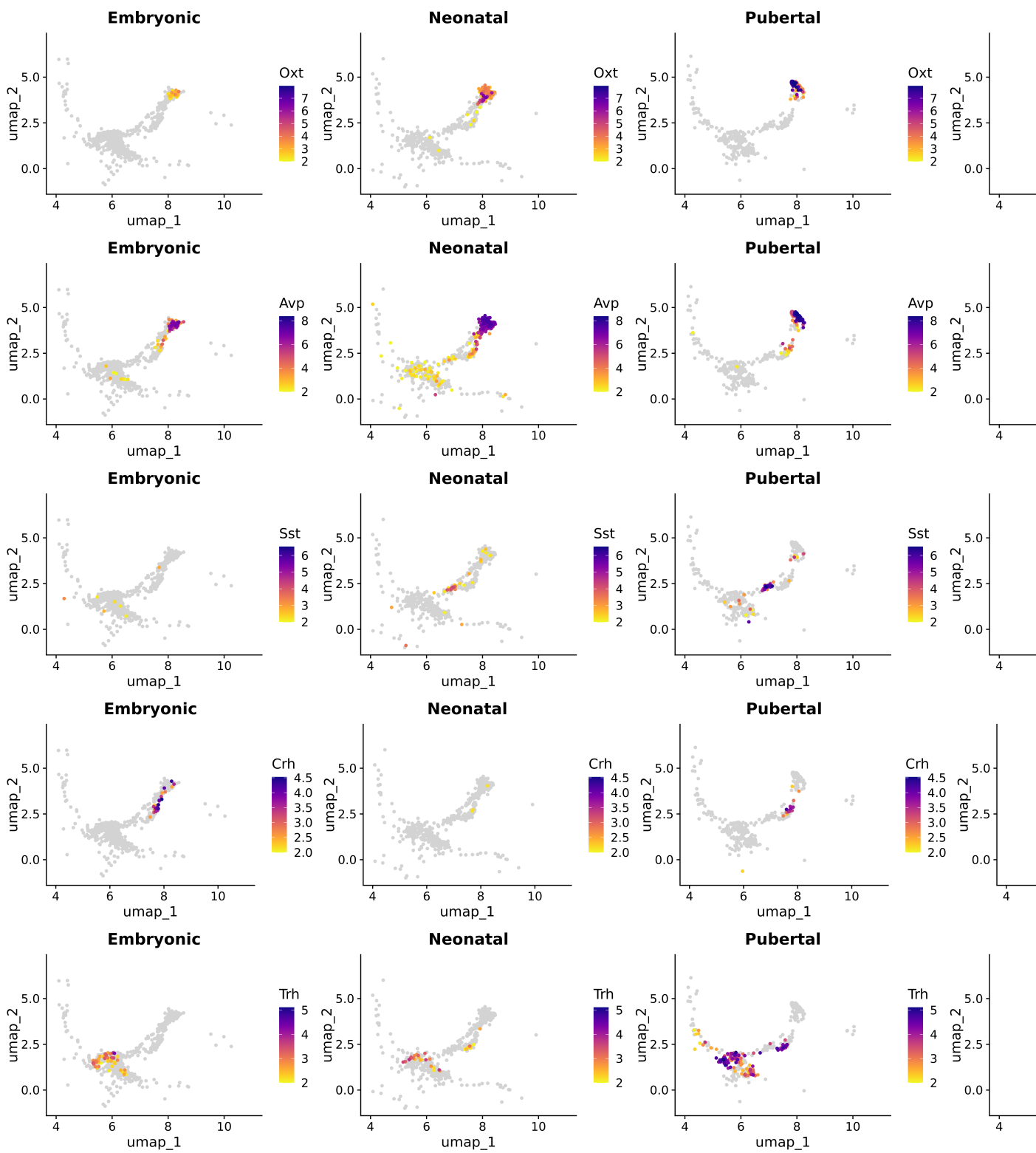
Adult

```
upset(
  as.data.frame(
    content_sbs_mtx_kim |>
      filter(stage == "Adult") |>
        select(
          c(cnbn) %>% [. %in% colnames(content_sbs_mtx_kim)]
        )
  ),
  order.by = "freq",
  group.by = "sets",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,
  point.size = 3.5, line.size = 2,
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
  nsets = 30,
  nintersects = 30,
  sets = c(cnbn) %>%
    [. %in% colnames(content_sbs_mtx_kim)],
  empty.intersections = F
)
```

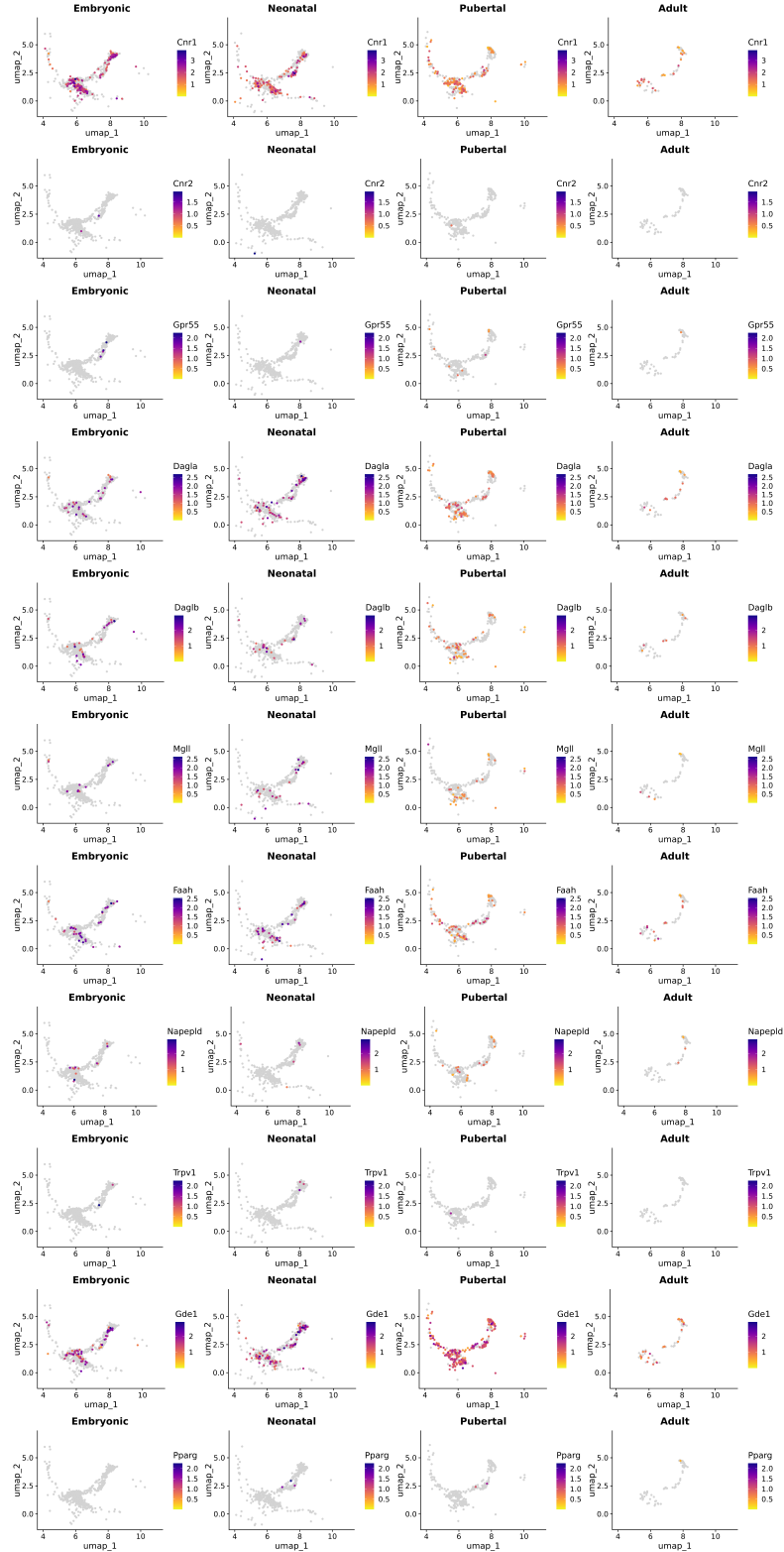


PVN Neurons from Romanov et al. 2020, Nature

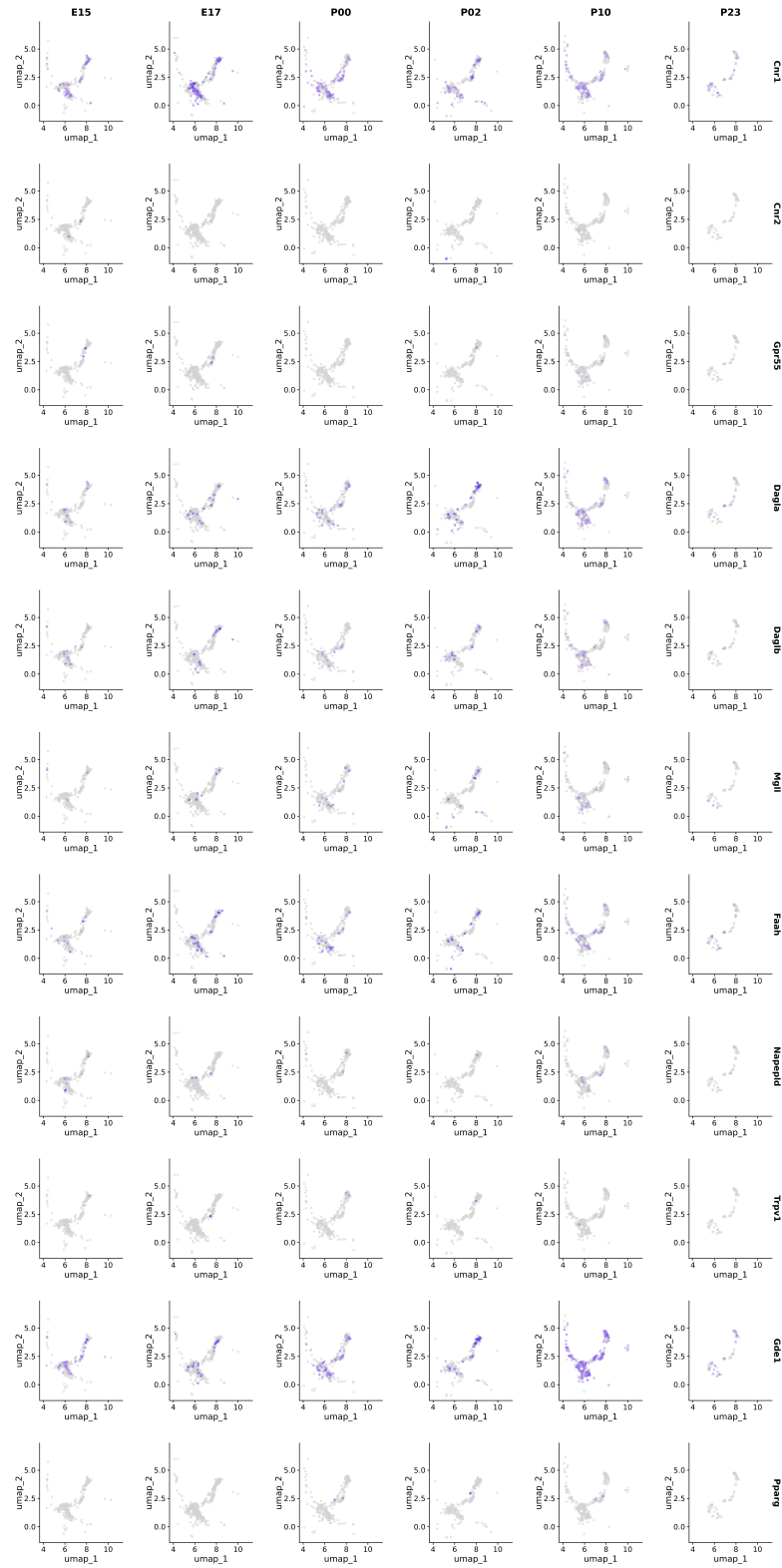
```
rar2020.srt.pvn <-  
  subset(  
    x = rar2020.srt.pub,  
    idents = c(  
      "mneOXY", "mneVAS",  
      "pneSS", "pneCRH", "pneTRH"  
    ),  
    invert = FALSE  
  )  
  
rar2020.srt.pvn <- subset(rar2020.srt.pvn, subset = umap_1 > 4 & umap_2 > -1)  
  
rar2020.srt.pvn$age <-  
  plyr::mapvalues(  
    x = rar2020.srt.pvn$age,  
    from = c("E15", "E17", "P0", "P2", "3P2", "1P10", "P10", "P23"),  
    to = c("E15", "E17", "P00", "P02", "P02", "P10", "P10", "P23")  
  )  
  
FeaturePlot_scCustom(  
  rar2020.srt.pvn,  
  reduction = "umap",  
  features = c(  
    "Oxt",  
    "Avp",  
    "Sst",  
    "Crh",  
    "Trh"),  
  split.by = "stage",  
  na_cutoff = 2,  
  label = F,  
  num_columns = 4  
) * NoLegend()
```



```
FeaturePlot_scCustom(rar2020.srt.pvn, reduction = "umap", features = c(cnb), split.
```



```
FeaturePlot(  
  rar2020.srt.pvn,  
  features = c(cnbn),  
  label = F,  
  blend = F,  
  order = TRUE,  
  pt.size = 1.2,  
  raster.dpi = c(1024, 1024),  
  alpha = 0.5,  
  split.by = "age"  
)
```




```

sbs_mtx <-
  rar2020.srt.pvn@assays$RNA@data %>%
  as.data.frame() %>%
  t()

# Filter features
filt_low_genes <-
  colSums(sbs_mtx) %>%
  .[, > quantile(., 0.4)] %>%
  names()
sbs_mtx %<>% .[, filt_low_genes]

min_filt_vector2 <-
  sbs_mtx %>%
  as_tibble() %>%
  select(all_of(filt_low_genes)) %>%
  summarise(across(.fns = ~ quantile(.x, .005))) %>%
  as.list() %>%
  map(as.double) %>%
  simplify() %>%
  .[filt_low_genes]

# Prepare table of intersection sets analysis
content_sbs_mtx_romanov <-
  (sbs_mtx > min_filt_vector2) %>%
  as_tibble() %>%
  mutate_all(as.numeric) %>%
  bind_cols(
    rar2020.srt.pvn@meta.data |> select(wtree, age, stage)
  )

```

All

```

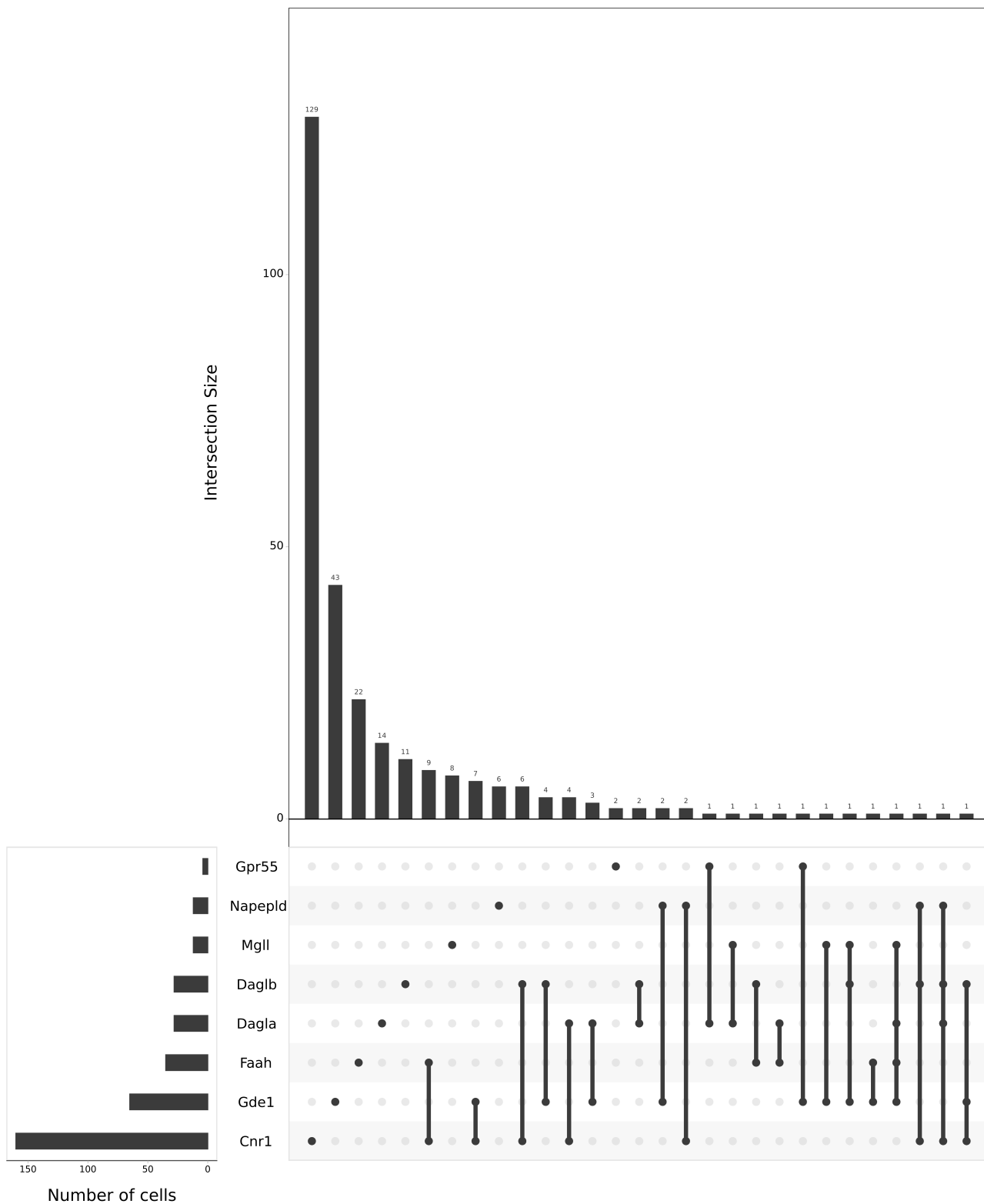
upset(
  as.data.frame(content_sbs_mtx_romanov),
  order.by = "freq",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,

```

```
point.size = 3.5, line.size = 2,  
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),  
nsets = 30,  
nintersects = 30,  
sets = c(cnbn) %>%  
  [. %in% colnames(content_sbs_mtx_romanov)],  
empty.intersections = NULL  
)
```

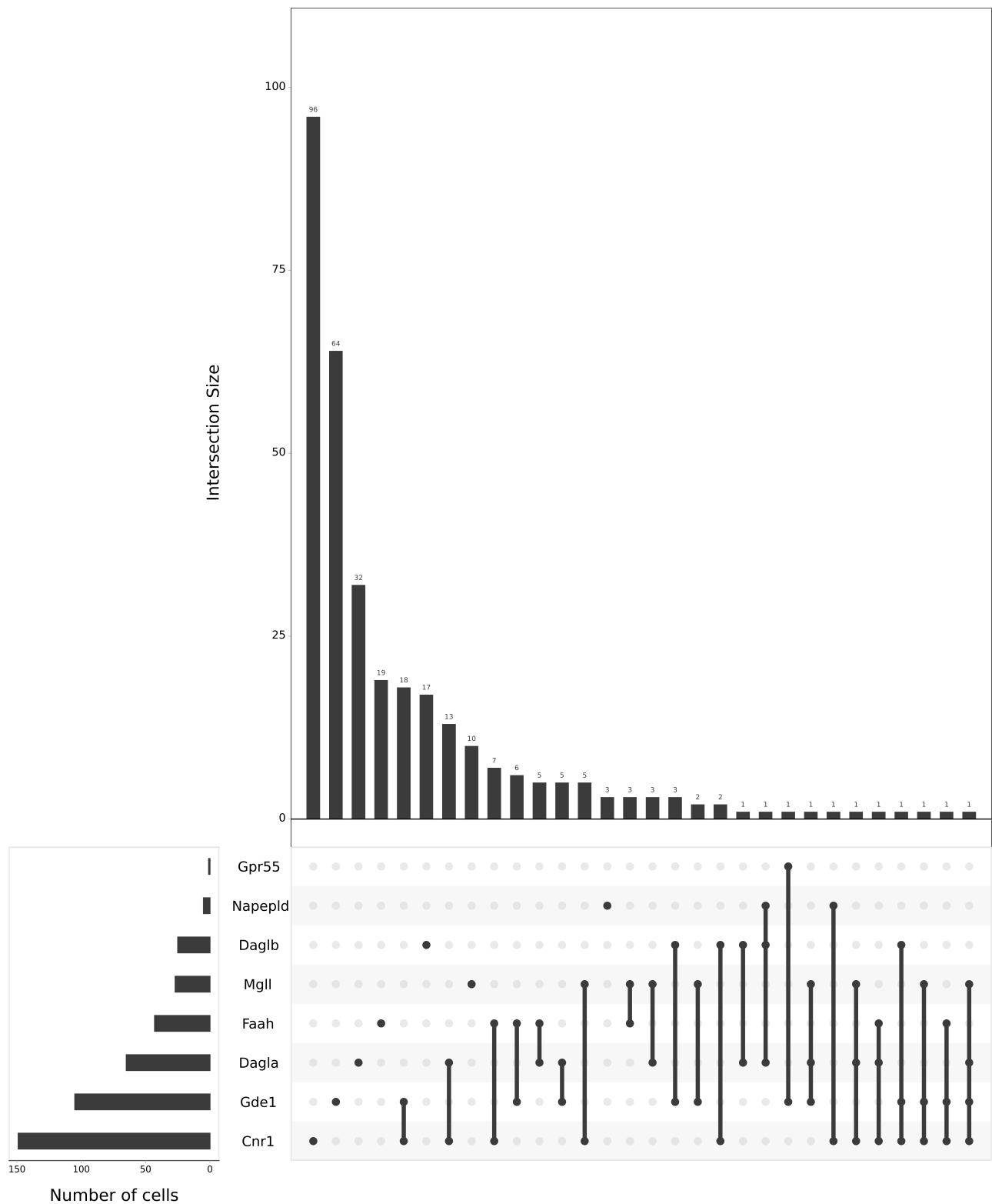

Embryonic

```
upset(
  as.data.frame(
    content_sbs_mtx_romanov |>
      filter(stage == "Embryonic") |>
        select(
          c(cnbn) %>% .[, %in% colnames(content_sbs_mtx_romanov)]
        )
  ),
  order.by = "freq",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,
  point.size = 3.5, line.size = 2,
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
  nsets = 30,
  nintersects = 30,
  sets = c(cnbn) %>%
    .[, %in% colnames(content_sbs_mtx_romanov)],
  empty.intersections = NULL
)
```



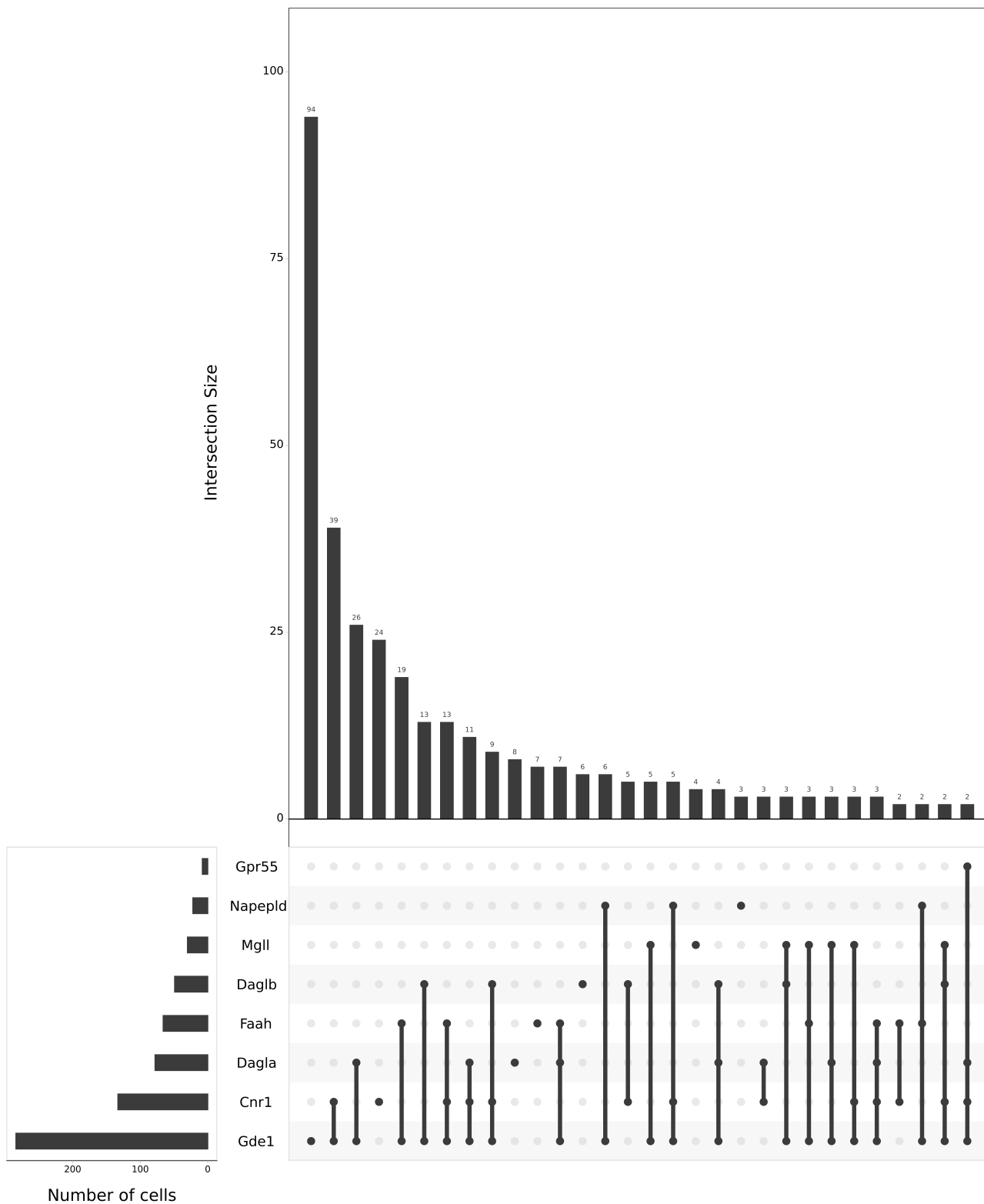
Neonatal

```
upset(  
  as.data.frame(  
    content_sbs_mtx_romanov |>  
    filter(stage == "Neonatal") |>  
    select(  
      c(cnbn) %>% .[. %in% colnames(content_sbs_mtx_romanov)]  
    )  
  ),  
  order.by = "freq",  
  cutoff = 3,  
  sets.x.label = "Number of cells",  
  number.angles = 0,  
  point.size = 3.5, line.size = 2,  
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),  
  nsets = 30,  
  nintersects = 30,  
  sets = c(cnbn) %>%  
    .[. %in% colnames(content_sbs_mtx_romanov)],  
  empty.intersections = NULL  
)
```



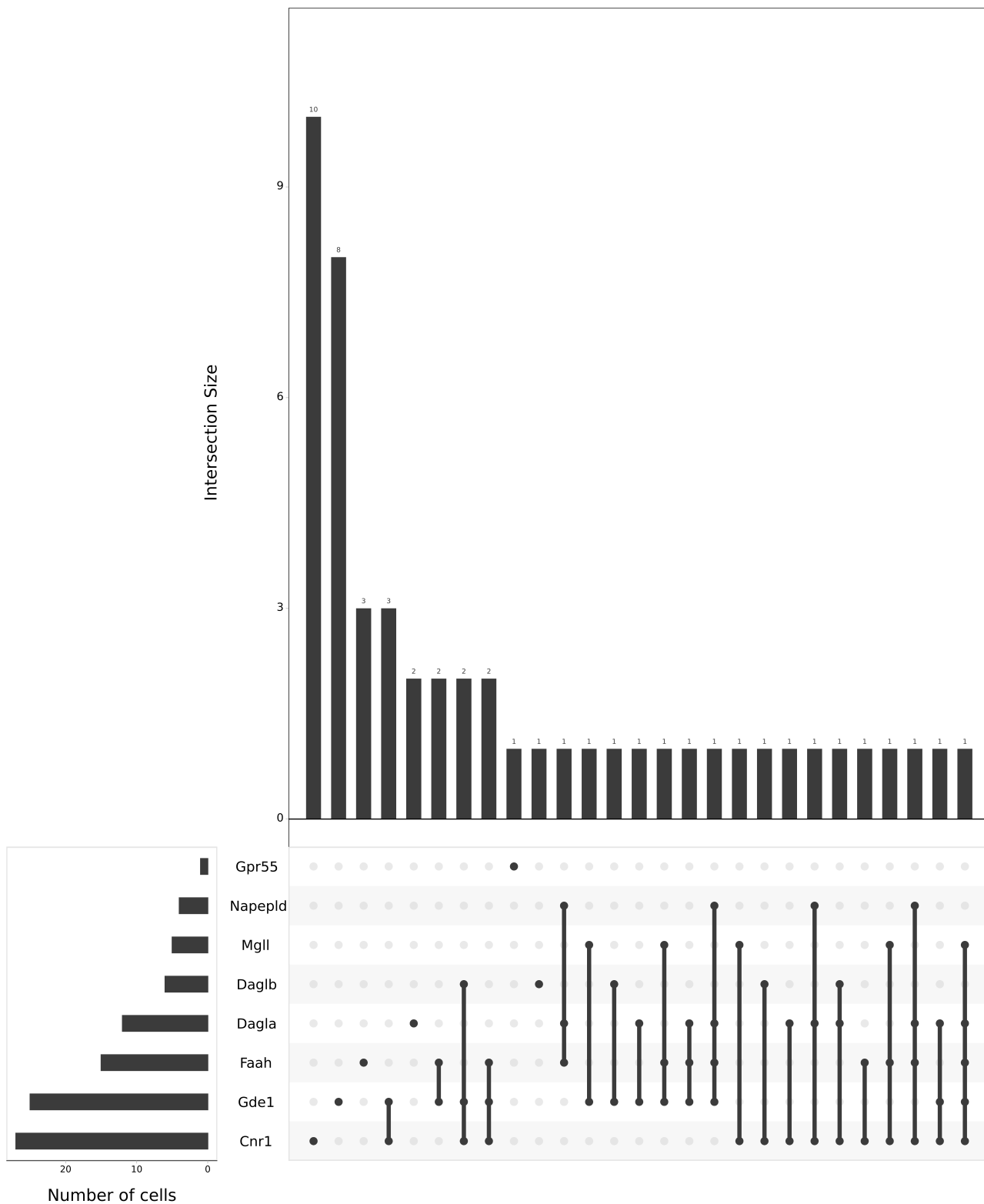
Pubertal

```
upset(  
  as.data.frame(  
    content_sbs_mtx_romanov |>  
    filter(stage == "Pubertal") |>  
    select(  
      c(cnbn) %>% .[. %in% colnames(content_sbs_mtx_romanov)]  
    )  
  ),  
  order.by = "freq",  
  cutoff = 3,  
  sets.x.label = "Number of cells",  
  number.angles = 0,  
  point.size = 3.5, line.size = 2,  
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),  
  nsets = 30,  
  nintersects = 30,  
  sets = c(cnbn) %>%  
    .[. %in% colnames(content_sbs_mtx_romanov)],  
  empty.intersections = NULL  
)
```

Adult

```
upset(
  as.data.frame(
    content_sbs_mtx_romanov |>
      filter(stage == "Adult") |>
        select(
          c(cnbn) %>% .[, %in% colnames(content_sbs_mtx_romanov)]
        )
  ),
  order.by = "freq",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,
  point.size = 3.5, line.size = 2,
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
  nsets = 30,
  nintersects = 30,
  sets = c(cnbn) %>%
    .[, %in% colnames(content_sbs_mtx_romanov)],
  empty.intersections = NULL
)
```



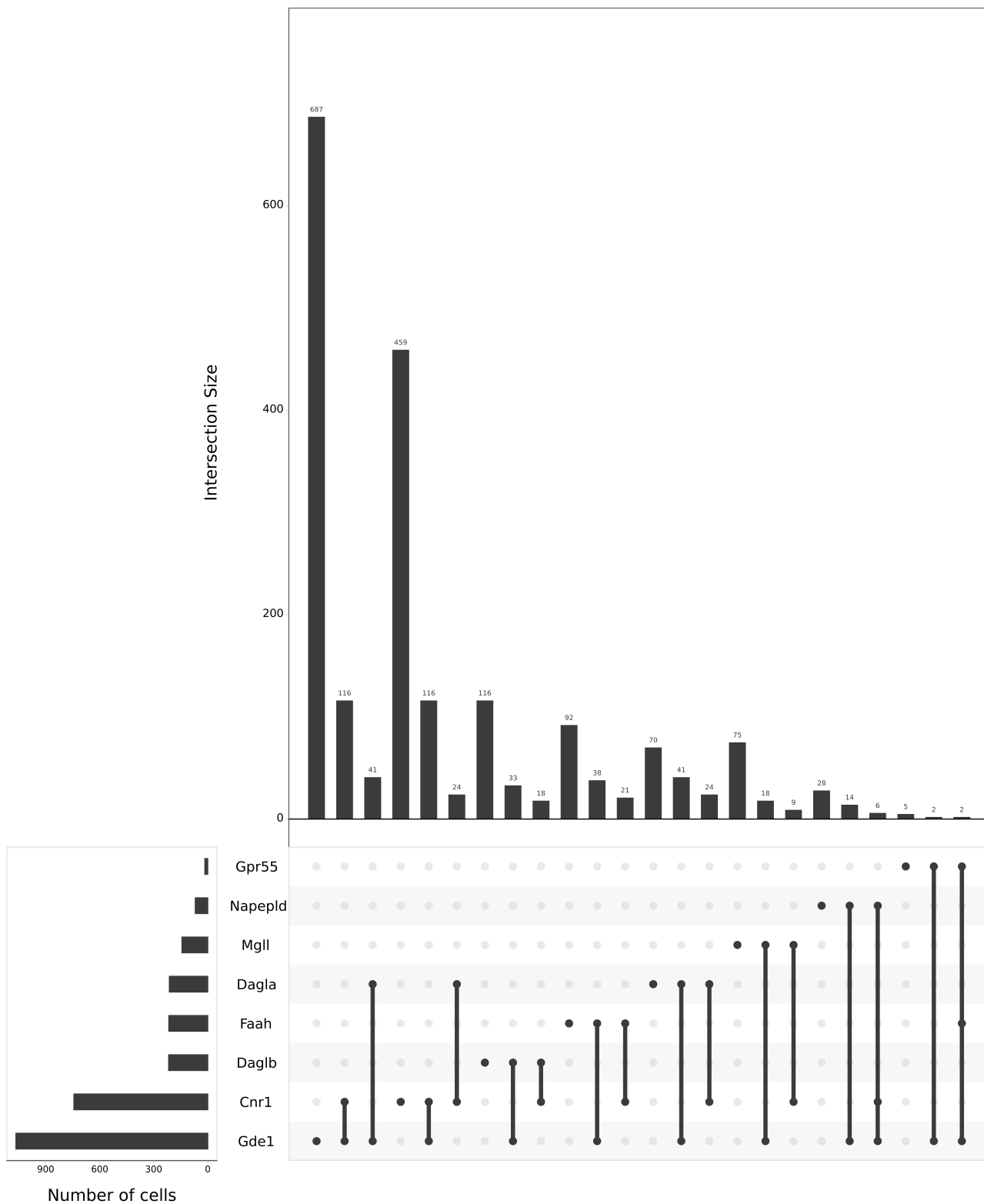
PVN Neurons from both datasets joined

```
# Prepare table of intersection sets analysis
to_select <-
  c(cnbn, "Oxt", "Crh", "Trh", "Avp", "Sst", "wtree", "age", "stage") %>%
  .[, %in% colnames(content_sbs_mtx_kim)] %>%
  .[, %in% colnames(content_sbs_mtx_romanov)]

content_sbs_mtx <-
  bind_rows(
    content_sbs_mtx_kim |> select(all_of(to_select)),
    content_sbs_mtx_romanov |> select(all_of(to_select))
  )
```

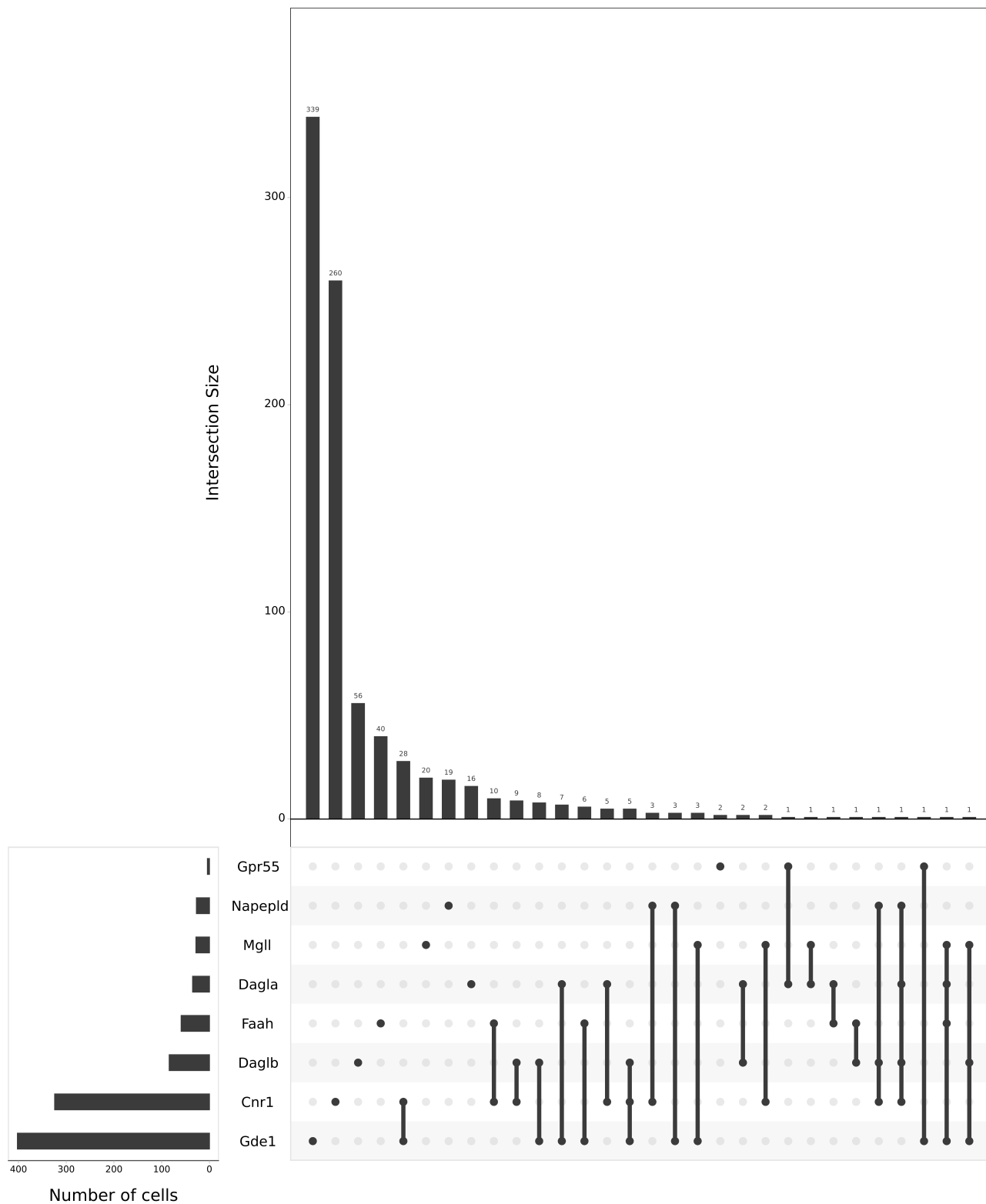
All

```
upset(
  as.data.frame(content_sbs_mtx),
  order.by = "freq",
  group.by = "sets",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,
  point.size = 3.5, line.size = 2,
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
  nsets = 30,
  nintersects = 30,
  sets = c(cnbn) %>%
    .[, %in% colnames(content_sbs_mtx)],
  empty.intersections = NULL
)
```



Embryonic

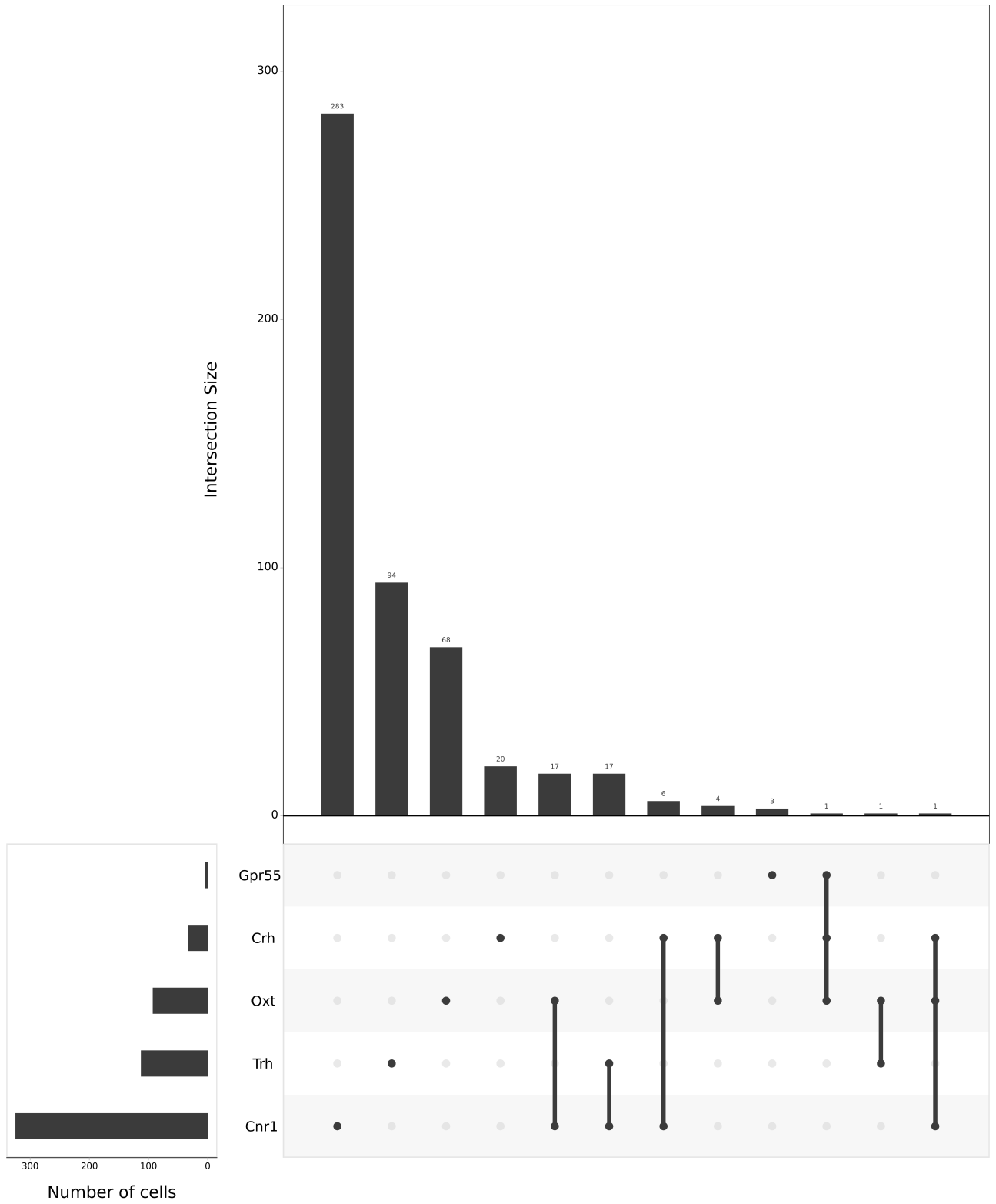
```
upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Embryonic") |>
      select(
        c(cnbn) %>% .[, %in% colnames(content_sbs_mtx)]
      )
  ),
  order.by = "freq",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,
  point.size = 3.5, line.size = 2,
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
  nsets = 30,
  nintersects = 30,
  sets = c(cnbn) %>%
    .[, %in% colnames(content_sbs_mtx)],
  empty.intersections = NULL
)
```



```

upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Embryonic") |>
      select(
        c("Cnr1", "Cnr2", "Gpr55", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(content_sbs_mtx)]
      )
  ),
  order.by = "freq",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,
  point.size = 3.5, line.size = 2,
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
  nsets = 30,
  nintersects = 30,
  sets = c("Cnr1", "Cnr2", "Gpr55", "Oxt", "Crh", "Trh") %>%
    .[, %in% colnames(content_sbs_mtx)],
  empty.intersections = NULL
)

```

```
skim(as.data.frame(
  content_sbs_mtx |>
    filter(stage == "Embryonic") |>
    select(
      c("Cnr1", "Cnr2", "Gpr55", "Oxt", "Crh", "Trh") %>% .[. %in% colnames(content_sbs_mtx)
    )
  ))
```

Table 1: Data summary

Name	as.data.frame(...)
Number of rows	3771
Number of columns	5
Column type frequency:	
numeric	5
Group variables	None

Variable type: numeric

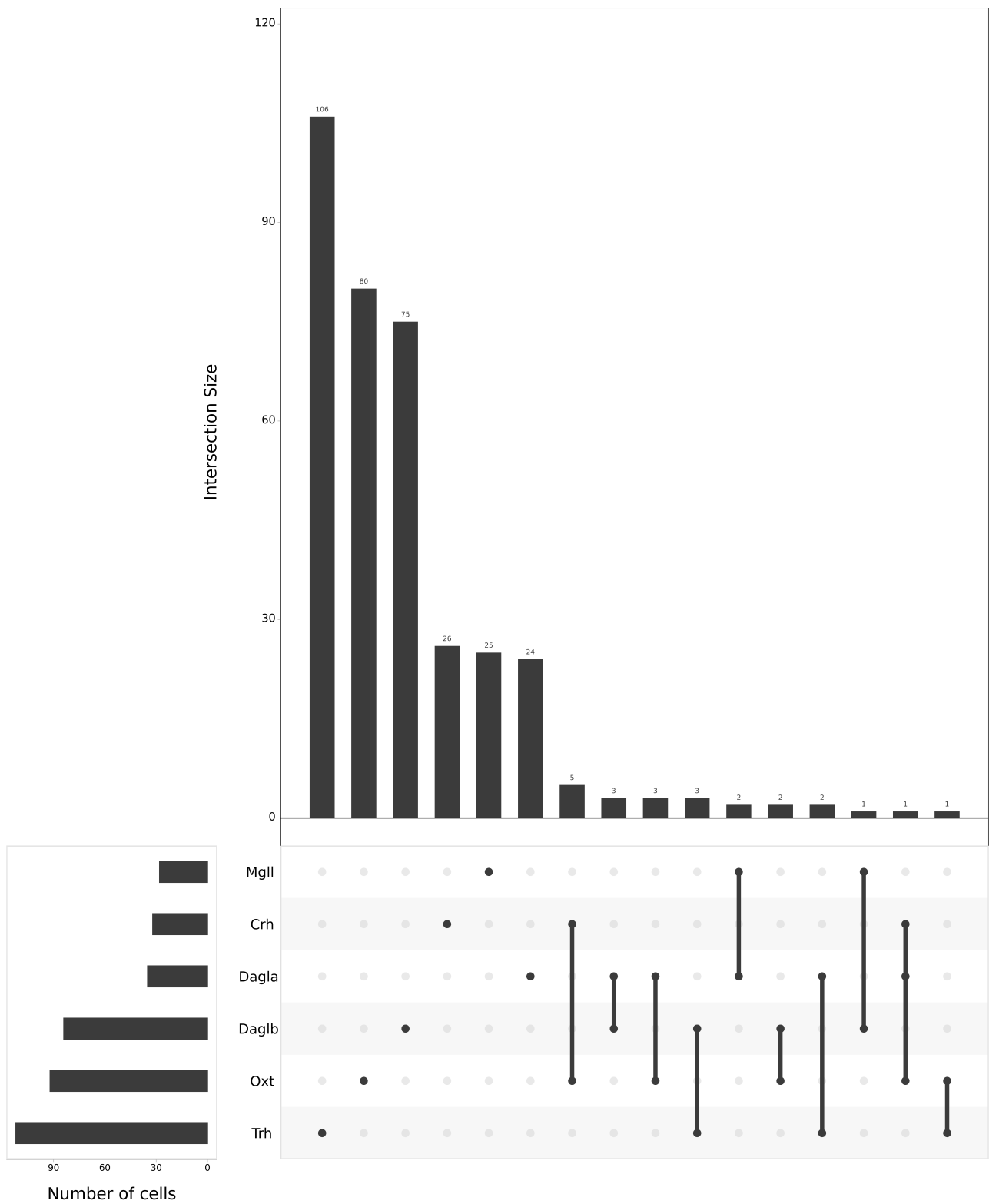
skim_variable	n_missing	complete	rate	mean	sd	p0	p25	p50	p75	p100	hist
Cnr1	0	1	0.09	0.28	0	0	0	0	0	1	
Gpr55	0	1	0.00	0.03	0	0	0	0	0	1	
Oxt	0	1	0.02	0.15	0	0	0	0	0	1	
Crh	0	1	0.01	0.09	0	0	0	0	0	1	
Trh	0	1	0.03	0.17	0	0	0	0	0	1	

```
upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Embryonic") |>
      select(
        c("Dagla", "Daglb", "Mgll", "Oxt", "Crh", "Trh") %>% .[. %in% colnames(content_sbs_mtx)
      )
  ),
  order.by = "freq",
  cutoff = 3,
```

```

sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c("Dagla", "Daglb", "Mgll", "Oxt", "Crh", "Trh") %>%
  [. %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

```



```
skim(as.data.frame(
  content_sbs_mtx |>
    filter(stage == "Embryonic") |>
    select(
      c("Dagla", "Daglb", "Mgll", "Oxt", "Crh", "Trh") %>% .[. %in% colnames(content_sbs_mtx)
    )
  ))
```

Table 3: Data summary

Name	as.data.frame(...)
Number of rows	3771
Number of columns	6
Column type frequency:	
numeric	6
Group variables	None

Variable type: numeric

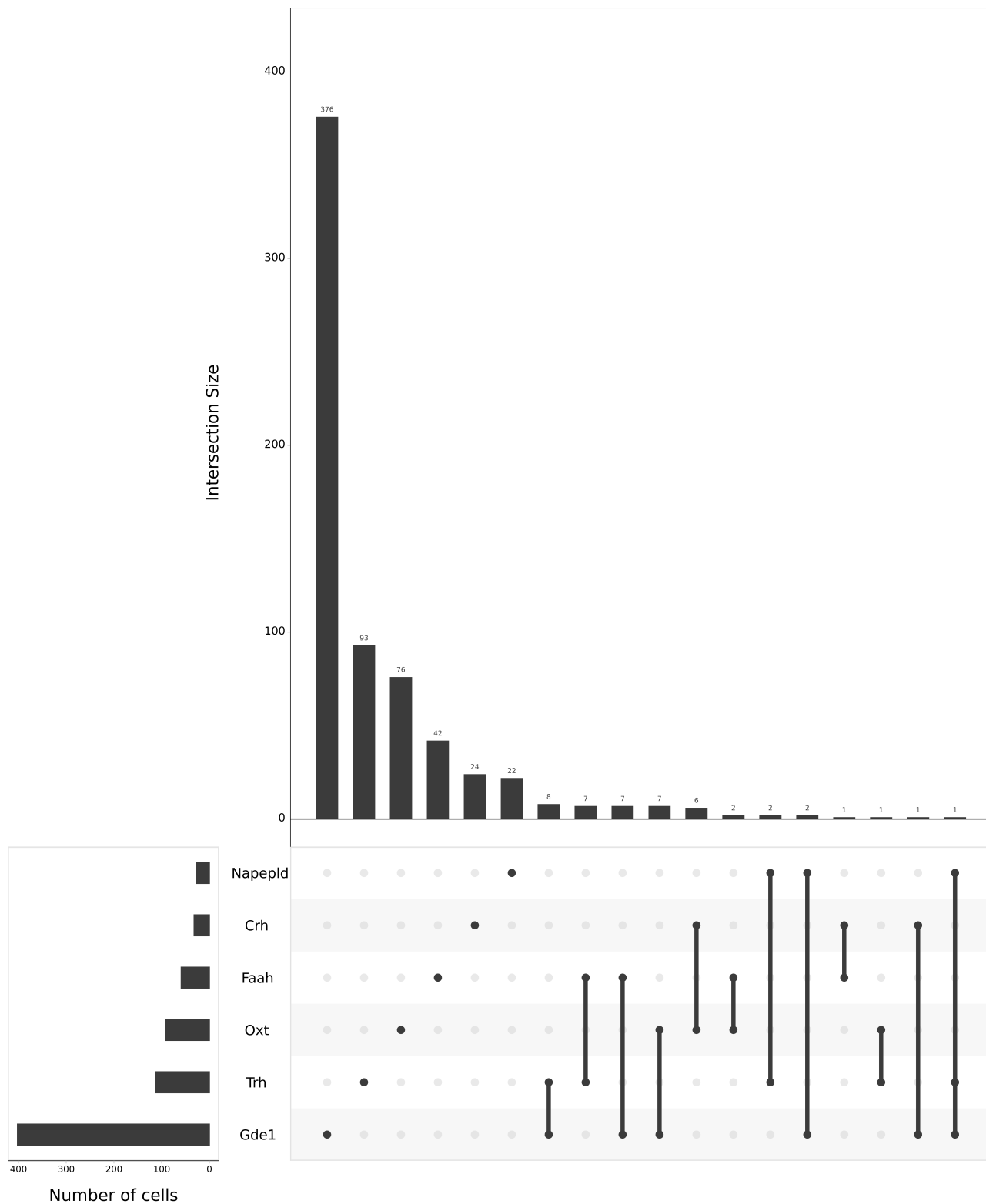
skim_	variablen_	missing	complete_	ratemean	sd	p0	p25	p50	p75	p100	hist
Dagla		0	1	0.01	0.10	0	0	0	0	1	
Daglb		0	1	0.02	0.15	0	0	0	0	1	
Mgll		0	1	0.01	0.09	0	0	0	0	1	
Oxt		0	1	0.02	0.15	0	0	0	0	1	
Crh		0	1	0.01	0.09	0	0	0	0	1	
Trh		0	1	0.03	0.17	0	0	0	0	1	

```
upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Embryonic") |>
      select(
        c("Napepld", "Gde1", "Faah", "Oxt", "Crh", "Trh") %>% .[. %in% colnames(content_sbs_mtx)
      )
  ),
  order.by = "freq",
```

```

cutoff = 3,
sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c("Napepld", "Gde1", "Faah", "Oxt", "Crh", "Trh") %>%
  [. %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

```



```
skim(as.data.frame(
  content_sbs_mtx |>
    filter(stage == "Embryonic") |>
    select(
      c("Napepld", "Gde1", "Faah", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(cont
    )
  ))
```

Table 5: Data summary

Name	as.data.frame(...)
Number of rows	3771
Number of columns	6
Column type frequency:	
numeric	6
Group variables	None

Variable type: numeric

skim_variabl	n_missing	complete	rate	mean	sd	p0	p25	p50	p75	p100	hist
Napepld	0	1	0.01	0.08	0	0	0	0	0	1	
Gde1	0	1	0.11	0.31	0	0	0	0	0	1	
Faah	0	1	0.02	0.12	0	0	0	0	0	1	
Oxt	0	1	0.02	0.15	0	0	0	0	0	1	
Crh	0	1	0.01	0.09	0	0	0	0	0	1	
Trh	0	1	0.03	0.17	0	0	0	0	0	1	

Neonatal

```
upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Neonatal") |>
      select(
        c(cnbn) %>% .[, %in% colnames(content_sbs_mtx)]
      )
  )
```



```

),
order.by = "freq",
cutoff = 3,
sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c(cnbn) %>%
  [. %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

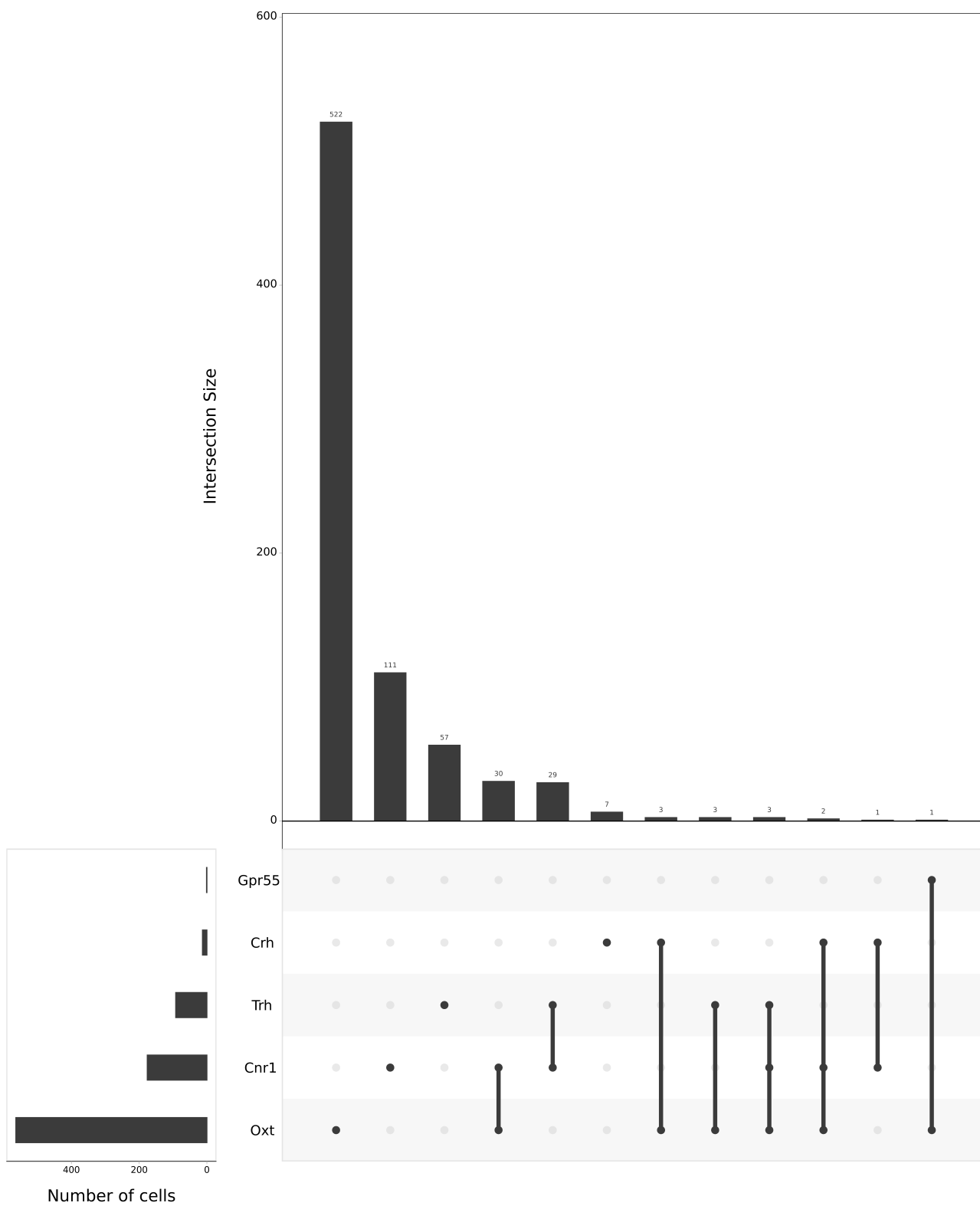
```



```

upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Neonatal") |>
      select(
        c("Cnr1", "Cnr2", "Gpr55", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(content_sbs_mtx)]
      )
  ),
  order.by = "freq",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,
  point.size = 3.5, line.size = 2,
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
  nsets = 30,
  nintersects = 30,
  sets = c("Cnr1", "Cnr2", "Gpr55", "Oxt", "Crh", "Trh") %>%
    .[, %in% colnames(content_sbs_mtx)],
  empty.intersections = NULL
)

```



```
skim(as.data.frame(
  content_sbs_mtx |>
    filter(stage == "Neonatal") |>
    select(
      c("Cnr1", "Cnr2", "Gpr55", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(content_sbs_mtx)]
    )
  ))
```

Table 7: Data summary

Name	as.data.frame(...)
Number of rows	1456
Number of columns	5
Column type frequency:	
numeric	5
Group variables	None

Variable type: numeric

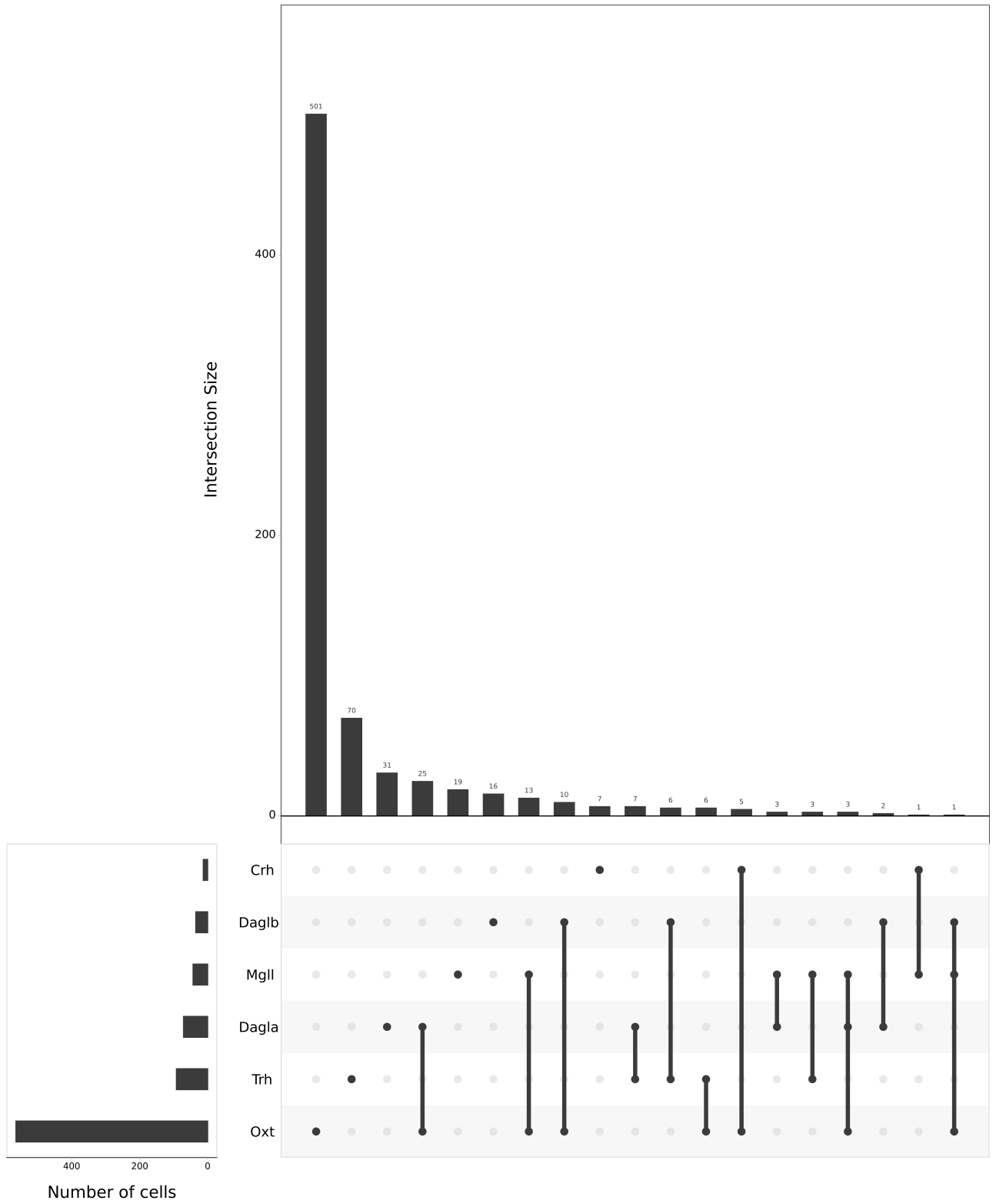
skim_vari	blen__missing	complete__rat	mean	sd	p0	p25	p50	p75	p100	hist
Cnr1	0	1	0.12	0.33	0	0	0	0	1	
Gpr55	0	1	0.00	0.03	0	0	0	0	1	
Oxt	0	1	0.39	0.49	0	0	0	1	1	
Crh	0	1	0.01	0.09	0	0	0	0	1	
Trh	0	1	0.06	0.24	0	0	0	0	1	

```
upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Neonatal") |>
      select(
        c("Dagla", "Daglb", "Mgll", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(content_sbs_mtx)]
      )
  ),
  order.by = "freq",
  cutoff = 3,
```

```

sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c("Dagla", "Daglb", "Mgll", "Oxt", "Crh", "Trh") %>%
  [. %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

```



```
skim(as.data.frame(
  content_sbs_mtx |>
    filter(stage == "Neonatal") |>
    select(
      c("Dagla", "Daglb", "Mgll", "Oxt", "Crh", "Trh") %>% .[. %in% colnames(content_sbs_mtx)
    )
  ))
```

Table 9: Data summary

Name	as.data.frame(...)
Number of rows	1456
Number of columns	6
Column type frequency:	
numeric	6
Group variables	None

Variable type: numeric

skim_variable	n_missing	complete	rate	mean	sd	p0	p25	p50	p75	p100	hist
Dagla	0	1	0.05	0.22	0	0	0	0	0	1	
Daglb	0	1	0.02	0.15	0	0	0	0	0	1	
Mgll	0	1	0.03	0.17	0	0	0	0	0	1	
Oxt	0	1	0.39	0.49	0	0	0	1	1	1	
Crh	0	1	0.01	0.09	0	0	0	0	0	1	
Trh	0	1	0.06	0.24	0	0	0	0	0	1	

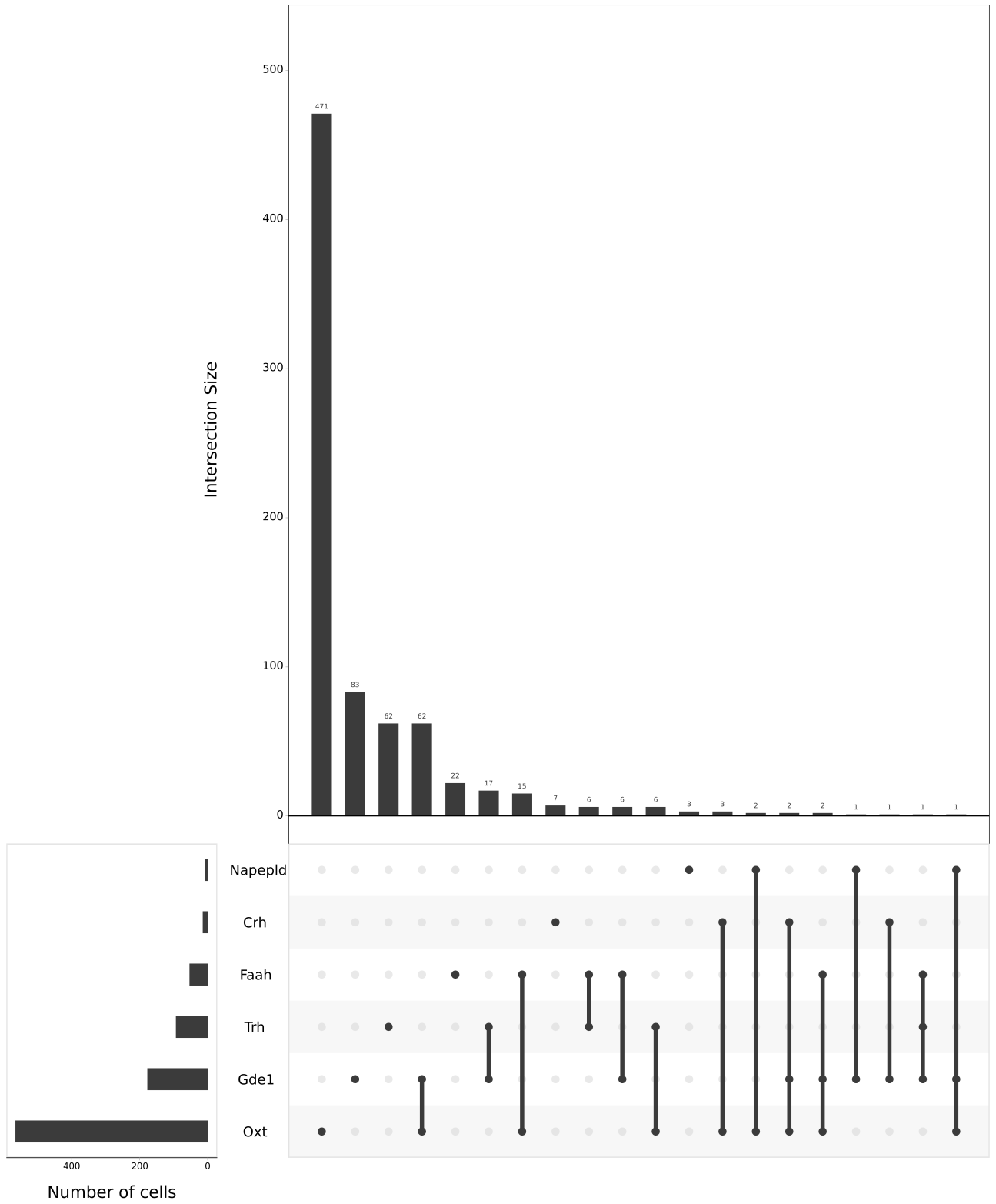
```
upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Neonatal") |>
      select(
        c("Napepld", "Gde1", "Faah", "Oxt", "Crh", "Trh") %>% .[. %in% colnames(content_sbs_mtx)
      )
  ),
  order.by = "freq",
```



```

cutoff = 3,
sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c("Napepld", "Gde1", "Faah", "Oxt", "Crh", "Trh") %>%
  [. %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

```



```
skim(as.data.frame(
  content_sbs_mtx |>
    filter(stage == "Neonatal") |>
    select(
      c("Napepld", "Gde1", "Faah", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(cont
    )
  ))
```

Table 11: Data summary

Name	as.data.frame(...)
Number of rows	1456
Number of columns	6
Column type frequency:	
numeric	6
Group variables	None

Variable type: numeric

skim_variabl	n_missing	complete	rate	mean	sd	p0	p25	p50	p75	p100	hist
Napepld	0	1	0.00	0.07	0	0	0	0	0	1	
Gde1	0	1	0.12	0.33	0	0	0	0	0	1	
Faah	0	1	0.04	0.19	0	0	0	0	0	1	
Oxt	0	1	0.39	0.49	0	0	0	1	1	1	
Crh	0	1	0.01	0.09	0	0	0	0	0	1	
Trh	0	1	0.06	0.24	0	0	0	0	0	1	

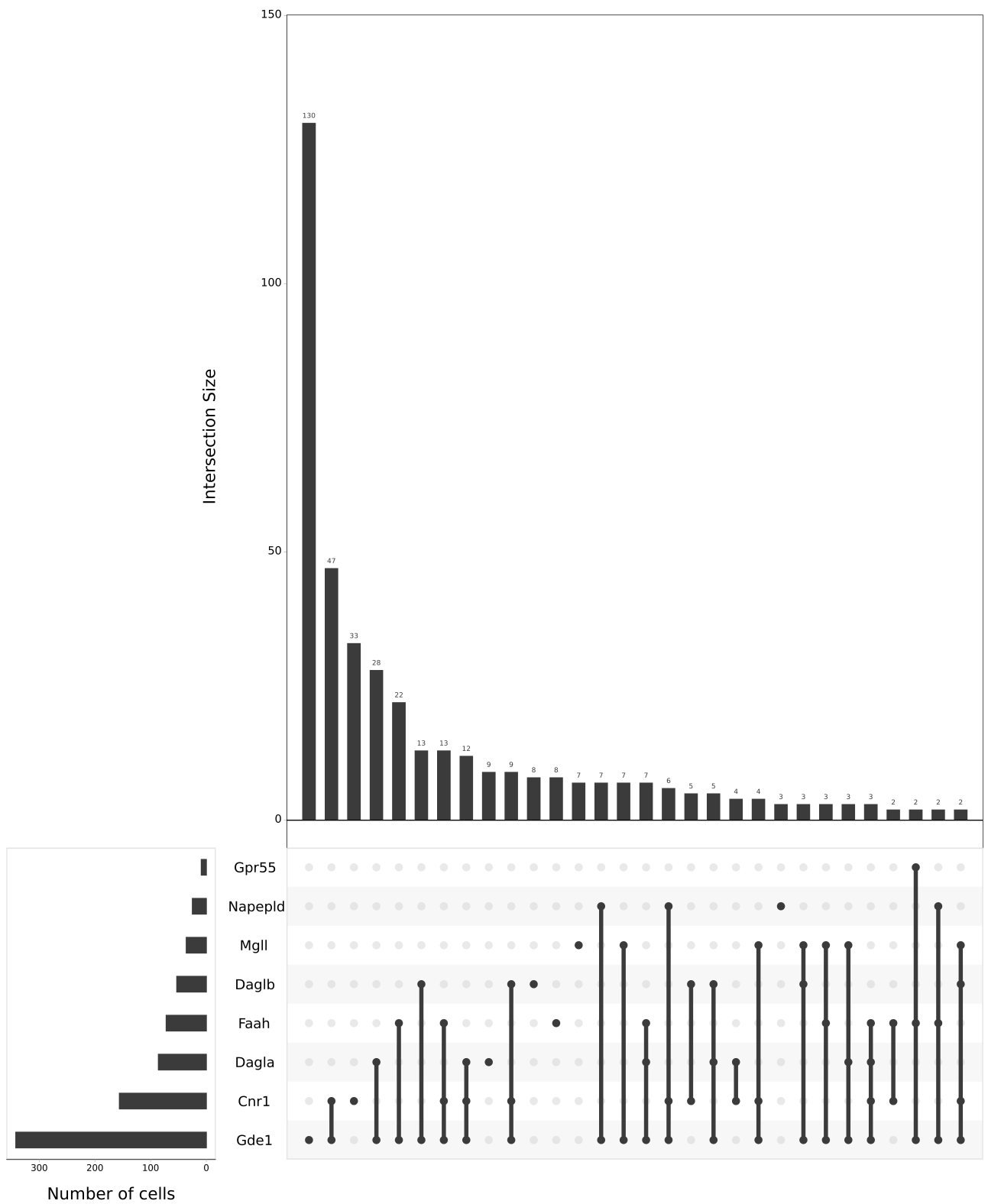
Pubertal

```
upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Pubertal") |>
      select(
        c(cnbn) %>% .[, %in% colnames(content_sbs_mtx)]
      )
  )
```

```

),
order.by = "freq",
cutoff = 3,
sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c(cnbn) %>%
  [. %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

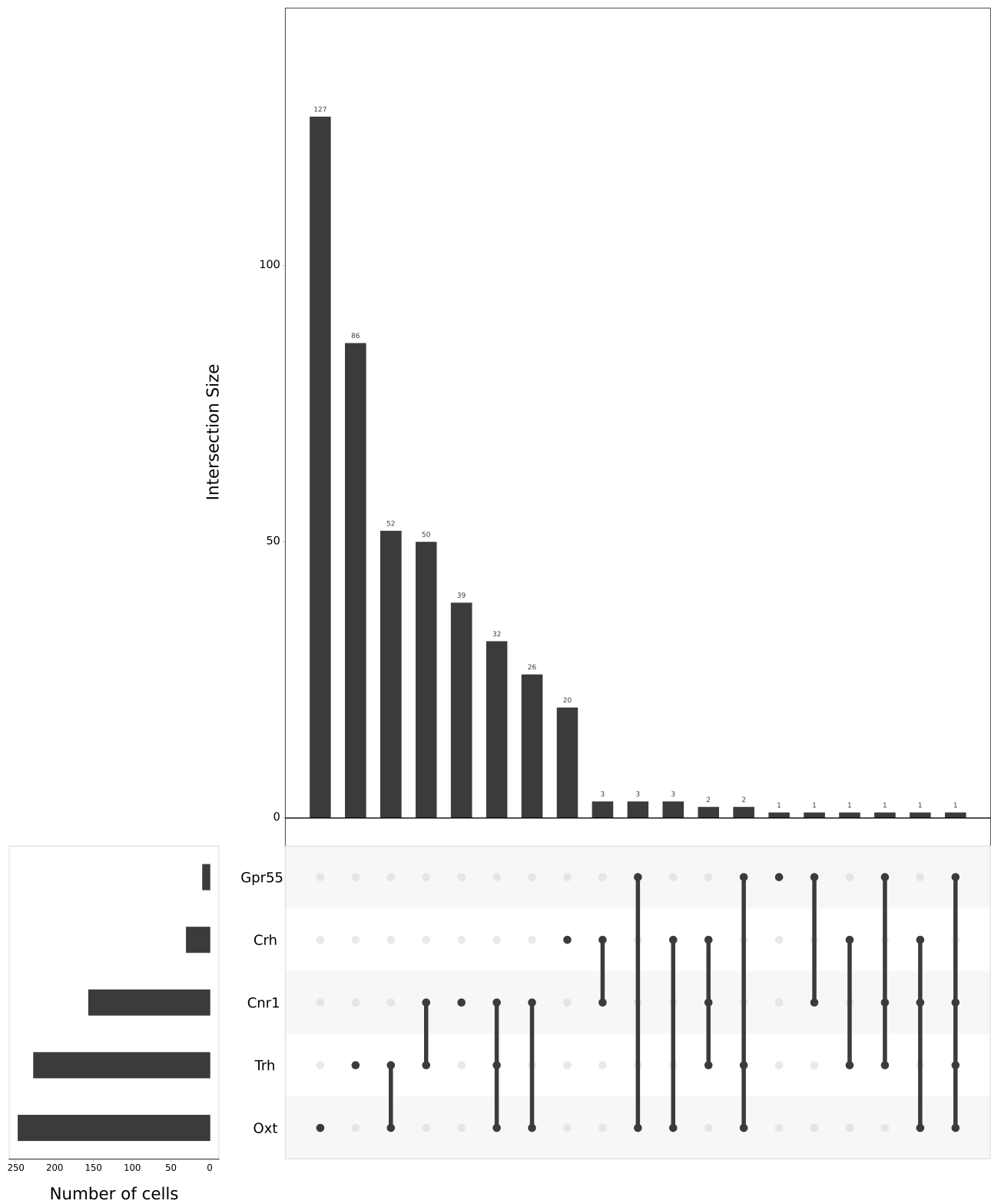
```



```

upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Pubertal") |>
      select(
        c("Cnr1", "Cnr2", "Gpr55", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(content_sbs_mtx)]
      )
  ),
  order.by = "freq",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,
  point.size = 3.5, line.size = 2,
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
  nsets = 30,
  nintersects = 30,
  sets = c("Cnr1", "Cnr2", "Gpr55", "Oxt", "Crh", "Trh") %>%
    .[, %in% colnames(content_sbs_mtx)],
  empty.intersections = NULL
)

```



```
skim(as.data.frame(
  content_sbs_mtx |>
    filter(stage == "Pubertal") |>
    select(
      c("Cnr1", "Cnr2", "Gpr55", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(content_sbs_mtx)]
    )
  ))
```

Table 13: Data summary

Name	as.data.frame(...)
Number of rows	641
Number of columns	5
Column type frequency:	
numeric	5
Group variables	None

Variable type: numeric

skim_variable	n_missing	complete	rate	mean	sd	p0	p25	p50	p75	p100	hist
Cnr1	0	1	0.24	0.43	0	0	0	0	0	1	
Gpr55	0	1	0.01	0.12	0	0	0	0	0	1	
Oxt	0	1	0.39	0.49	0	0	0	1	1	1	
Crh	0	1	0.05	0.21	0	0	0	0	0	1	
Trh	0	1	0.35	0.48	0	0	0	1	1	1	

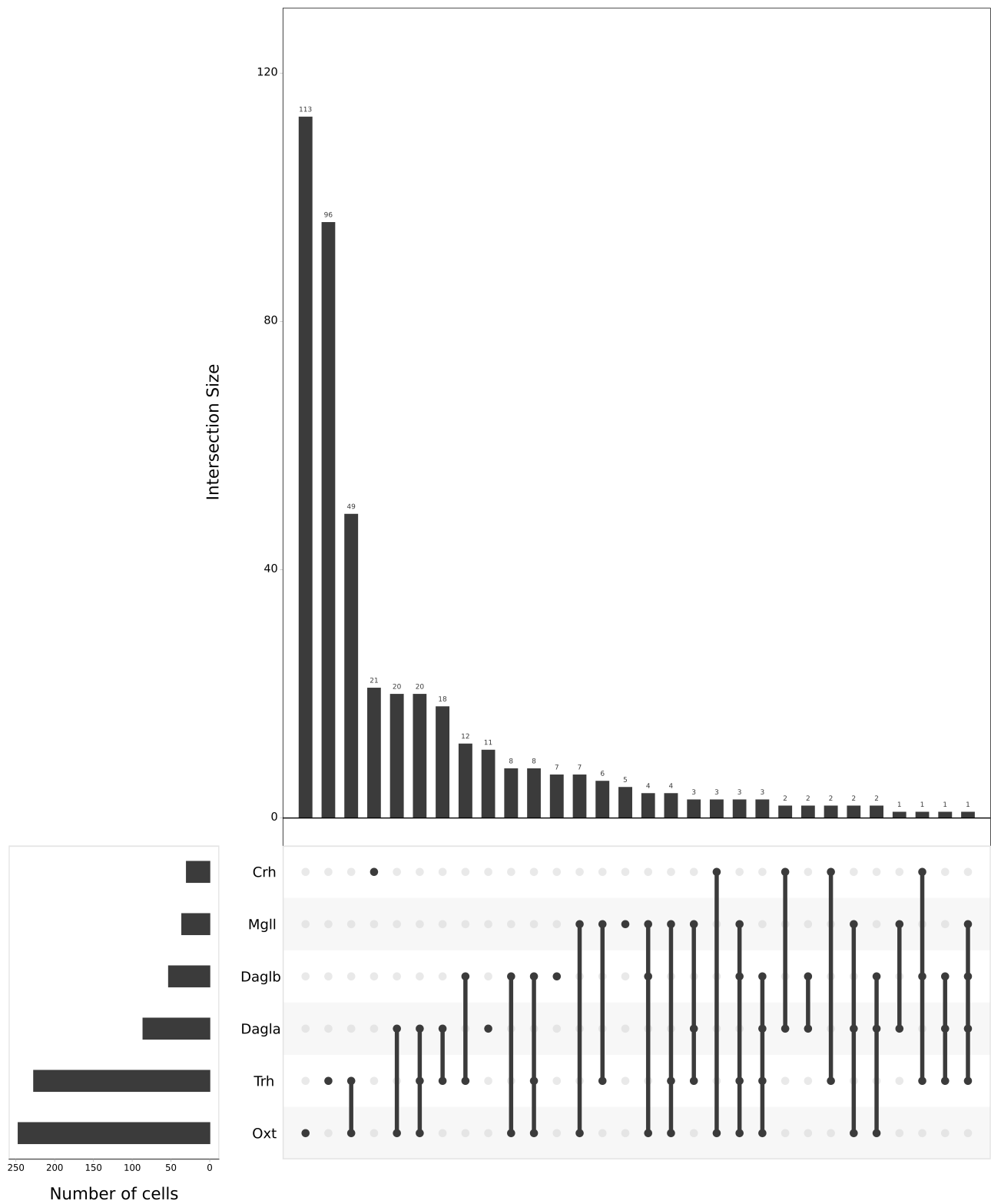
```
upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Pubertal") |>
      select(
        c("Dagla", "Daglb", "Mgll", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(content_sbs_mtx)]
      )
  ),
  order.by = "freq",
  cutoff = 3,
```



```

sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c("Dagla", "Daglb", "Mgll", "Oxt", "Crh", "Trh") %>%
  [. %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

```



```
skim(as.data.frame(
  content_sbs_mtx |>
    filter(stage == "Pubertal") |>
    select(
      c("Dagla", "Daglb", "Mgll", "Oxt", "Crh", "Trh") %>% .[. %in% colnames(content_sbs_mtx)
    )
  ))
```

Table 15: Data summary

Name	as.data.frame(...)
Number of rows	641
Number of columns	6
Column type frequency:	
numeric	6
Group variables	None

Variable type: numeric

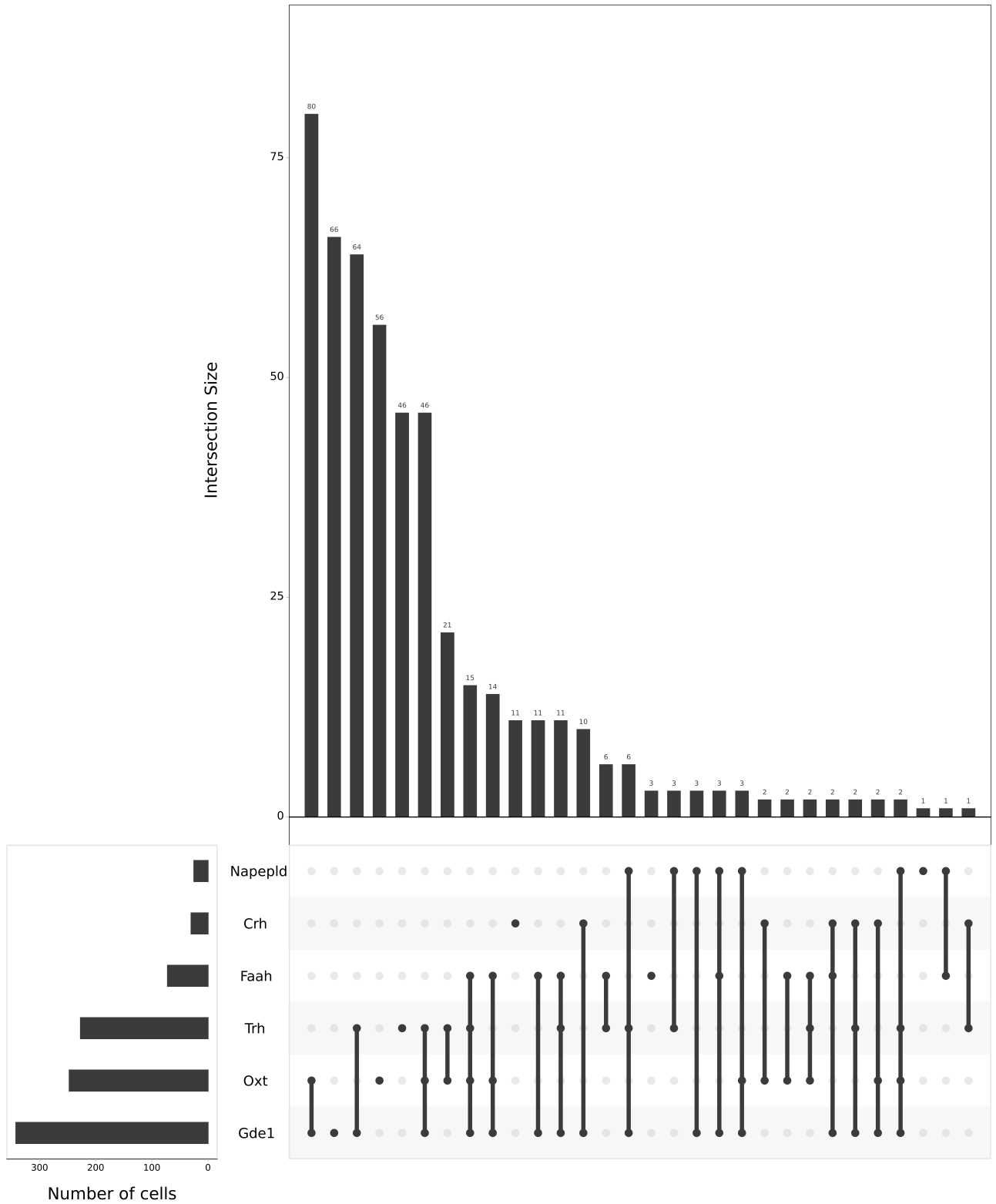
skim_vari	blen	_missing	complete	_rate	mean	sd	p0	p25	p50	p75	p100	hist
Dagla		0	1	0.13	0.34	0	0	0	0	0	1	
Daglb		0	1	0.08	0.28	0	0	0	0	0	1	
Mgll		0	1	0.06	0.23	0	0	0	0	0	1	
Oxt		0	1	0.39	0.49	0	0	0	1	1	1	
Crh		0	1	0.05	0.21	0	0	0	0	0	1	
Trh		0	1	0.35	0.48	0	0	0	1	1	1	

```
upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Pubertal") |>
      select(
        c("Napepld", "Gde1", "Faah", "Oxt", "Crh", "Trh") %>% .[. %in% colnames(content_sbs_mtx)
      )
  ),
  order.by = "freq",
```

```

cutoff = 3,
sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c("Napepld", "Gde1", "Faah", "Oxt", "Crh", "Trh") %>%
  [. %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

```



```
skim(as.data.frame(
  content_sbs_mtx |>
    filter(stage == "Pubertal") |>
    select(
      c("Napepld", "Gde1", "Faah", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(cont
    )
  ))
```

Table 17: Data summary

Name	as.data.frame(...)
Number of rows	641
Number of columns	6
Column type frequency:	
numeric	6
Group variables	None

Variable type: numeric

skim_variabl	n_missing	complete	rate	mean	sd	p0	p25	p50	p75	p100	hist
Napepld	0	1	0.04	0.19	0	0	0	0	0	1	
Gde1	0	1	0.53	0.50	0	0	1	1	1	1	
Faah	0	1	0.11	0.32	0	0	0	0	0	1	
Oxt	0	1	0.39	0.49	0	0	0	1	1	1	
Crh	0	1	0.05	0.21	0	0	0	0	0	1	
Trh	0	1	0.35	0.48	0	0	0	1	1	1	

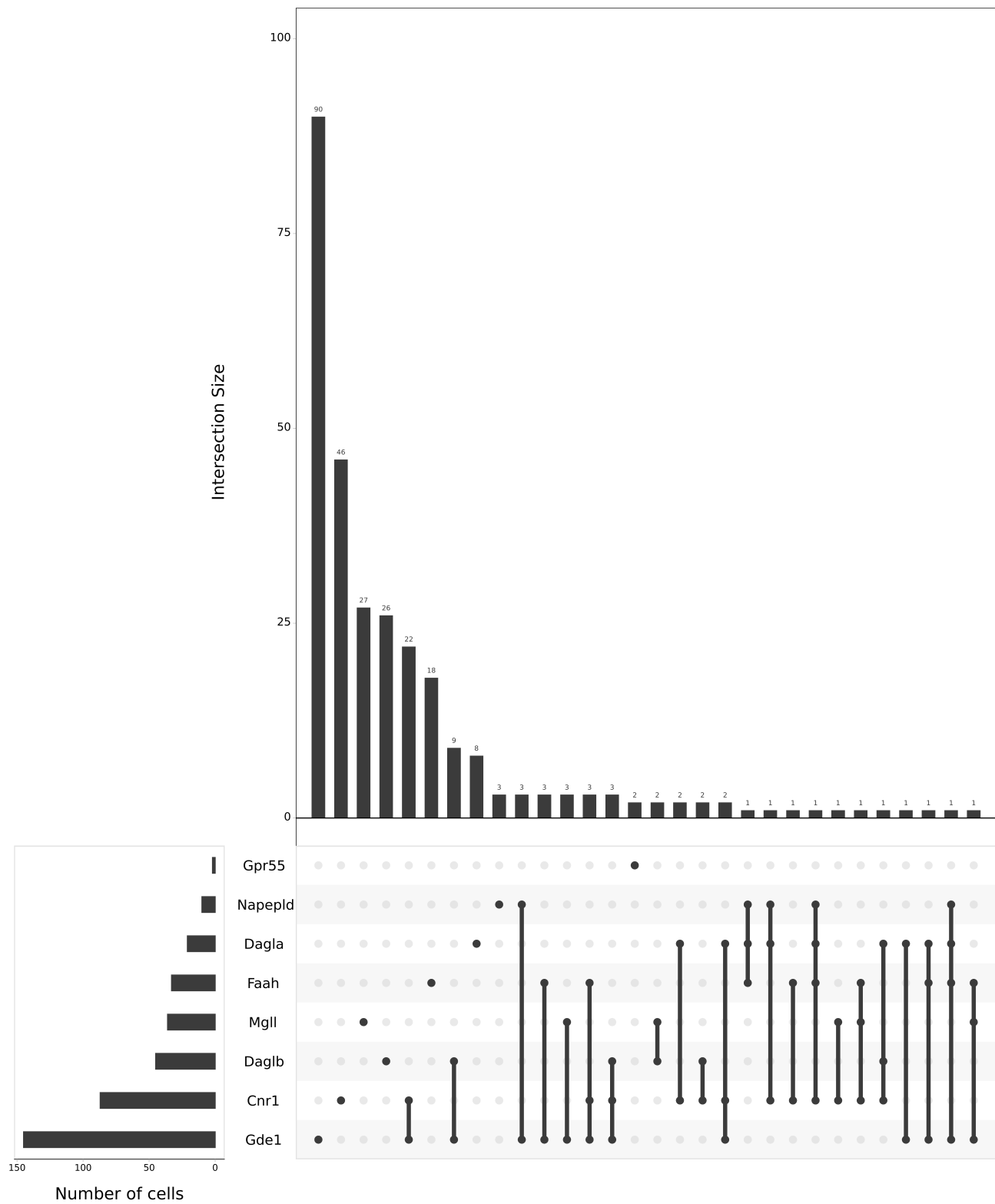
Adult

```
upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Adult") |>
      select(
        c(cnbn) %>% .[, %in% colnames(content_sbs_mtx)]
      )
  )
```

```

),
order.by = "freq",
cutoff = 3,
sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c(cnbn) %>%
  .[, %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

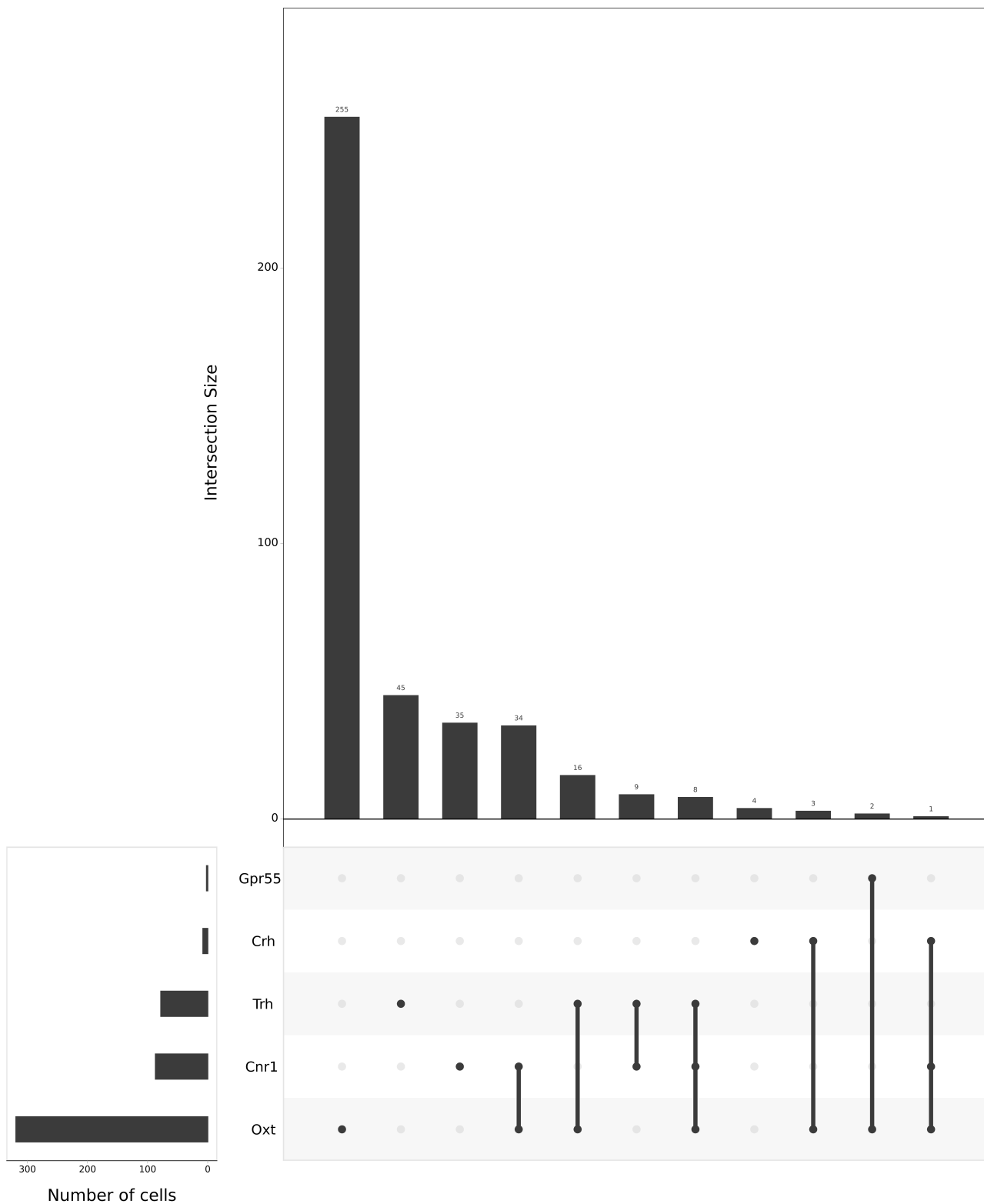
```




```

upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Adult") |>
      select(
        c("Cnr1", "Cnr2", "Gpr55", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(content_sbs_mtx)]
      )
  ),
  order.by = "freq",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,
  point.size = 3.5, line.size = 2,
  text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
  nsets = 30,
  nintersects = 30,
  sets = c("Cnr1", "Cnr2", "Gpr55", "Oxt", "Crh", "Trh") %>%
    .[, %in% colnames(content_sbs_mtx)],
  empty.intersections = NULL
)

```



```
skim(as.data.frame(
  content_sbs_mtx |>
    filter(stage == "Adult") |>
    select(
      c("Cnr1", "Cnr2", "Gpr55", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(content_sbs_mtx)]
    )
  ))
```

Table 19: Data summary

Name	as.data.frame(...)
Number of rows	801
Number of columns	5
Column type frequency:	
numeric	5
Group variables	None

Variable type: numeric

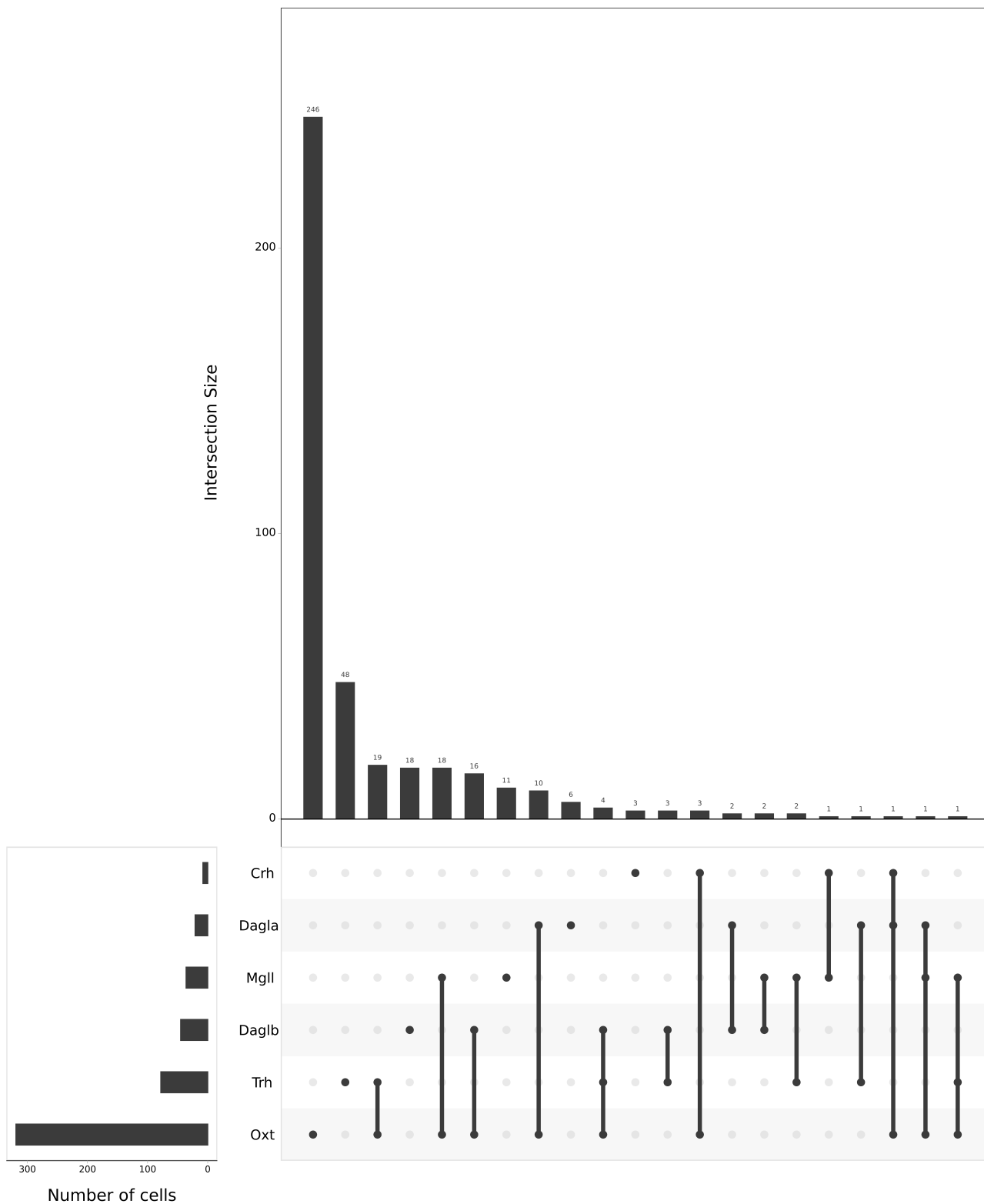
skim_variable	n_missing	complete	rate	mean	sd	p0	p25	p50	p75	p100	hist
Cnr1	0	1	0.11	0.31	0	0	0	0	0	1	
Gpr55	0	1	0.00	0.05	0	0	0	0	0	1	
Oxt	0	1	0.40	0.49	0	0	0	1	1	1	
Crh	0	1	0.01	0.10	0	0	0	0	0	1	
Trh	0	1	0.10	0.30	0	0	0	0	0	1	

```
upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Adult") |>
      select(
        c("Dagla", "Daglb", "Mgll", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(content_sbs_mtx)]
      )
  ),
  order.by = "freq",
  cutoff = 3,
```

```

sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c("Dagla", "Daglb", "Mgll", "Oxt", "Crh", "Trh") %>%
  [. %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

```



```
skim(as.data.frame(
  content_sbs_mtx |>
    filter(stage == "Adult") |>
    select(
      c("Dagla", "Daglb", "Mgll", "Oxt", "Crh", "Trh") %>% .[. %in% colnames(content_sbs_mtx)
    )
  ))
```

Table 21: Data summary

Name	as.data.frame(...)
Number of rows	801
Number of columns	6
Column type frequency:	
numeric	6
Group variables	None

Variable type: numeric

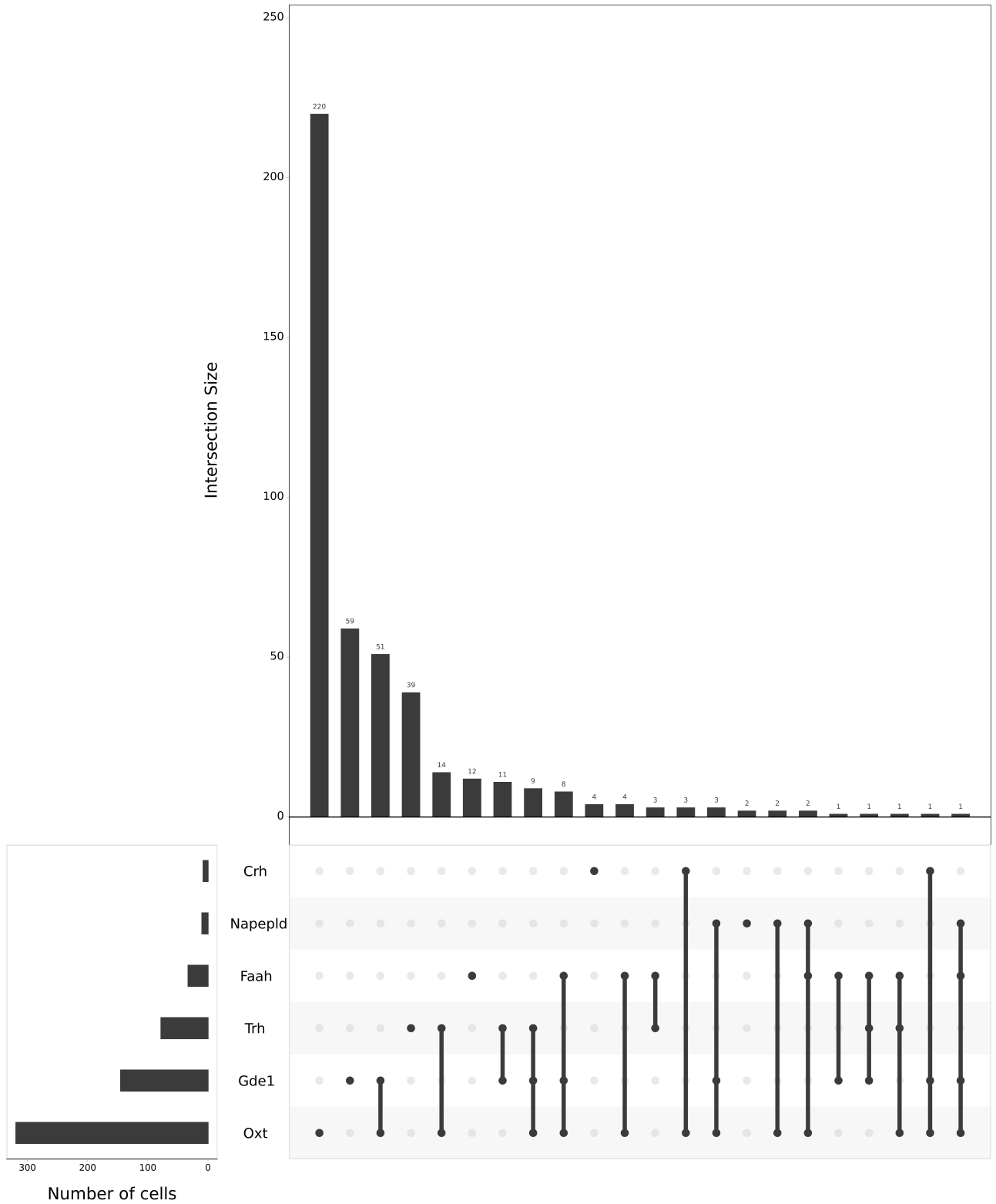
skim_variable	n_missing	complete	rate	mean	sd	p0	p25	p50	p75	p100	hist
Dagla	0	1	0.03	0.16	0	0	0	0	0	1	
Daglb	0	1	0.06	0.23	0	0	0	0	0	1	
Mgll	0	1	0.04	0.21	0	0	0	0	0	1	
Oxt	0	1	0.40	0.49	0	0	0	1	1	1	
Crh	0	1	0.01	0.10	0	0	0	0	0	1	
Trh	0	1	0.10	0.30	0	0	0	0	0	1	

```
upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(stage == "Adult") |>
      select(
        c("Napepld", "Gde1", "Faah", "Oxt", "Crh", "Trh") %>% .[. %in% colnames(content_sbs_mtx)
      )
  ),
  order.by = "freq",
```

```

cutoff = 3,
sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c("Napepld", "Gde1", "Faah", "Oxt", "Crh", "Trh") %>%
  [. %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

```




```
skim(as.data.frame(
  content_sbs_mtx |>
    filter(stage == "Adult") |>
    select(
      c("Napepld", "Gde1", "Faah", "Oxt", "Crh", "Trh") %>% .[, %in% colnames(cont
    )
  ))
```

Table 23: Data summary

Name	as.data.frame(...)
Number of rows	801
Number of columns	6
Column type frequency:	
numeric	6
Group variables	None

Variable type: numeric

skim_variabl	n_missing	complete	rate	mean	sd	p0	p25	p50	p75	p100	hist
Napepld	0	1	0.01	0.11	0	0	0	0	0	1	
Gde1	0	1	0.18	0.39	0	0	0	0	0	1	
Faah	0	1	0.04	0.20	0	0	0	0	0	1	
Oxt	0	1	0.40	0.49	0	0	0	1	1	1	
Crh	0	1	0.01	0.10	0	0	0	0	0	1	
Trh	0	1	0.10	0.30	0	0	0	0	0	1	

Contingency tables

```
rar2020.srt.pvn$age %>% forcats::fct_count()
```

```
# A tibble: 6 x 2
  f      n
  <fct> <int>
1 E15    317
```

2	E17	593
3	P00	350
4	P02	362
5	P10	425
6	P23	67

```
rar2020.srt.pvn$stage %>% forcats::fct_count()
```

```
# A tibble: 4 x 2
  f           n
  <fct>     <int>
1 Embryonic  910
2 Neonatal   712
3 Pubertal   425
4 Adult       67
```

```
srt$age %>% forcats::fct_count()
```

```
# A tibble: 12 x 2
  f           n
  <fct>     <int>
1 E10        22
2 E11       168
3 E12       374
4 E13       290
5 E14       540
6 E15       444
7 E16       655
8 E18       368
9 P14       216
10 P4        199
11 P45       734
12 P8       545
```

```
srt$stage %>% forcats::fct_count()
```

```
# A tibble: 4 x 2
  f      n
<fct> <int>
1 Embryonic 2861
2 Neonatal 744
3 Pubertal 216
4 Adult 734
```

Load selected astrocytes data from Lopez JP et al (2021)

```
anndata <- sc$read(here(
  "../1_heteroAstrocytes/PRJNA679294/data/",
  "class_cello/PRJNA679294-whole_dataset-0.001-cello_annotation.h5ad"
))
```

Convert adata object to R AnnDataR6 object.

```
adata <- py_to_r(anndata)
class(adata)
```

```
[1] "AnnDataR6" "R6"
```

```
class(adata$X)
```

```
[1] "dgCMatrix"
attr(,"package")
[1] "Matrix"
```

```
adata
```

AnnData object with n_obs × n_vars = 9572 × 22835

```
obs: 'nCount_RAW', 'nFeature_RAW', 'nCount_RNA', 'nFeature_RNA', 'orig.ident', 'nF
var: 'vst.mean', 'vst.variance', 'vst.variance.expected', 'vst.variance.standardiz
uns: 'k_tree_colors', 'name', 'ora_celltype_colors'
obsm: 'X_pacmap', 'X_pca', 'X_umap', 'ora_estimate', 'ora_pvals'
```

```

expr_mtx <- t(as.matrix(adata$raw$X))
colnames(expr_mtx) <- rownames(adata$X)
rownames(expr_mtx) <- adata$var_names
srt <- CreateSeuratObject(
  expr_mtx,
  assay = "RNA",
  project = "individual_hypothalamic_nuclei_astrocytes_evaluation_dataset",
  meta.data = as.data.frame(adata$obs)
)

Idents(srt) <- "ora_celltype"
srt <- subset(srt, idents = c("Astrocytes"))

Idents(srt) <- "libname"

print(srt)

```

An object of class Seurat
 22835 features across 2838 samples within 1 assay
 Active assay: RNA (22835 features, 0 variable features)
 1 layer present: counts

```

rm(adata, anndata, expr_mtx)
invisible(gc())

```

```

srt <- NormalizeData(srt)
srt <- FindVariableFeatures(srt, selection.method = "vst", nfeatures = 5000)
all.genes <- rownames(srt)
srt <- ScaleData(srt, features = c(VariableFeatures(srt), cnbn))

```

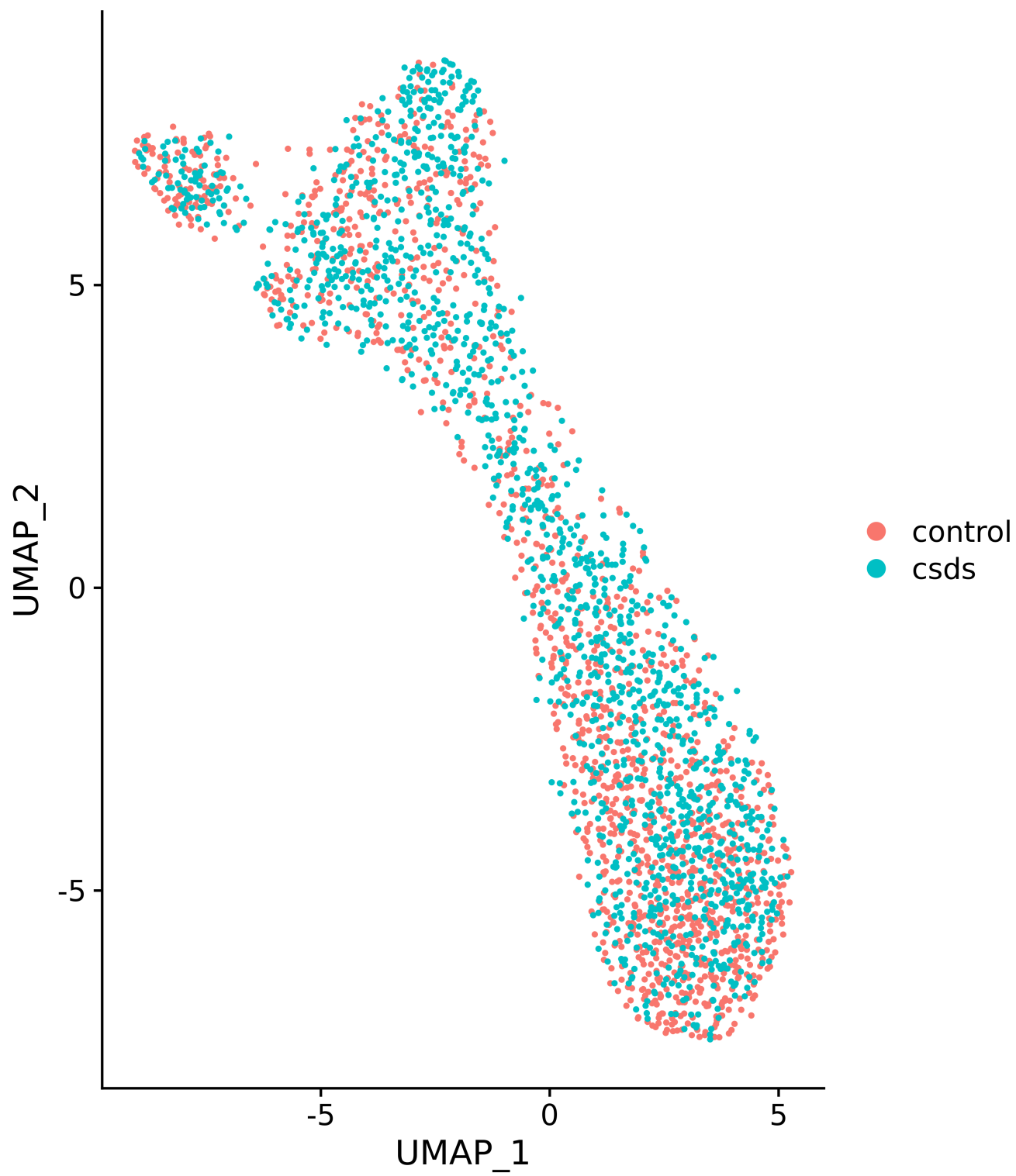
```

srt <- RunPCA(srt, npcs = 50, verbose = FALSE)
srt <-
  srt |>
  FindNeighbors(
    dims = 1:40,
    k.param = 40,
    annoy.metric = "cosine",
    n.trees = 100,

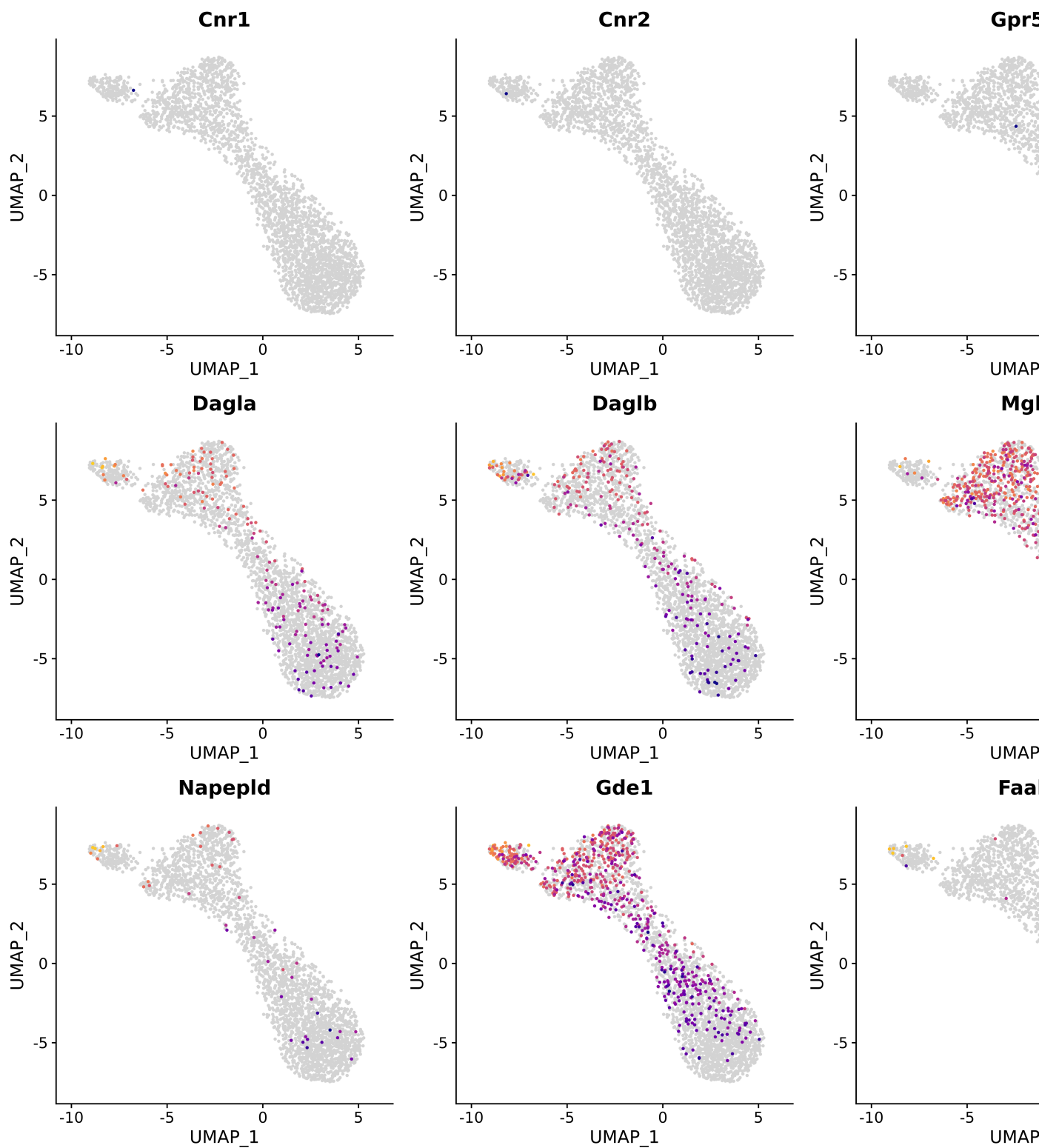
```

```
    verbose = FALSE
) |>
RunUMAP(
  dims = 1:50,
  reduction.name = "umap",
  reduction.key = "UMAP_",
  return.model = F,
  n.epochs = 1000L,
  n.neighbors = 50,
  min.dist = 0.5,
  metric = "cosine",
  seed.use = reseed,
  verbose = FALSE
)
```

```
DimPlot(srt)
```



```
FeaturePlot_scCustom(  
  srt,  
  reduction = "umap",  
  features = c(  
    "Cnr1", "Cnr2", "Gpr55", "Slc1a3",  
    "Dagla", "Daglb", "Mgll", "Gfap",  
    "Napepld", "Gde1", "Faah", "Aldh1l1"  
  ),  
  label = F,  
  num_columns = 4  
) * NoLegend()
```




```

sbs_mtx <-
  srt@assays$RNA@layers$data %>%
  as.data.frame() %>%
  t()

rownames(sbs_mtx) <- colnames(srt)
colnames(sbs_mtx) <- rownames(srt)

# Filter features
filt_low_genes <-
  colSums(sbs_mtx) %>%
  .[, > quantile(., 0.4)] %>%
  names()
sbs_mtx %<>% .[, filt_low_genes]

min_filt_vector2 <-
  sbs_mtx %>%
  as_tibble() %>%
  select(all_of(filt_low_genes)) %>%
  summarise(across(.fns = ~ quantile(.x, .005))) %>%
  as.list() %>%
  map(as.double) %>%
  simplify() %>%
  .[filt_low_genes]

# Prepare table of intersection sets analysis
content_sbs_mtx <-
  (sbs_mtx > min_filt_vector2) %>%
  as_tibble() %>%
  mutate_all(as.numeric) %>%
  bind_cols(
    srt@meta.data |> select(condit)
  )

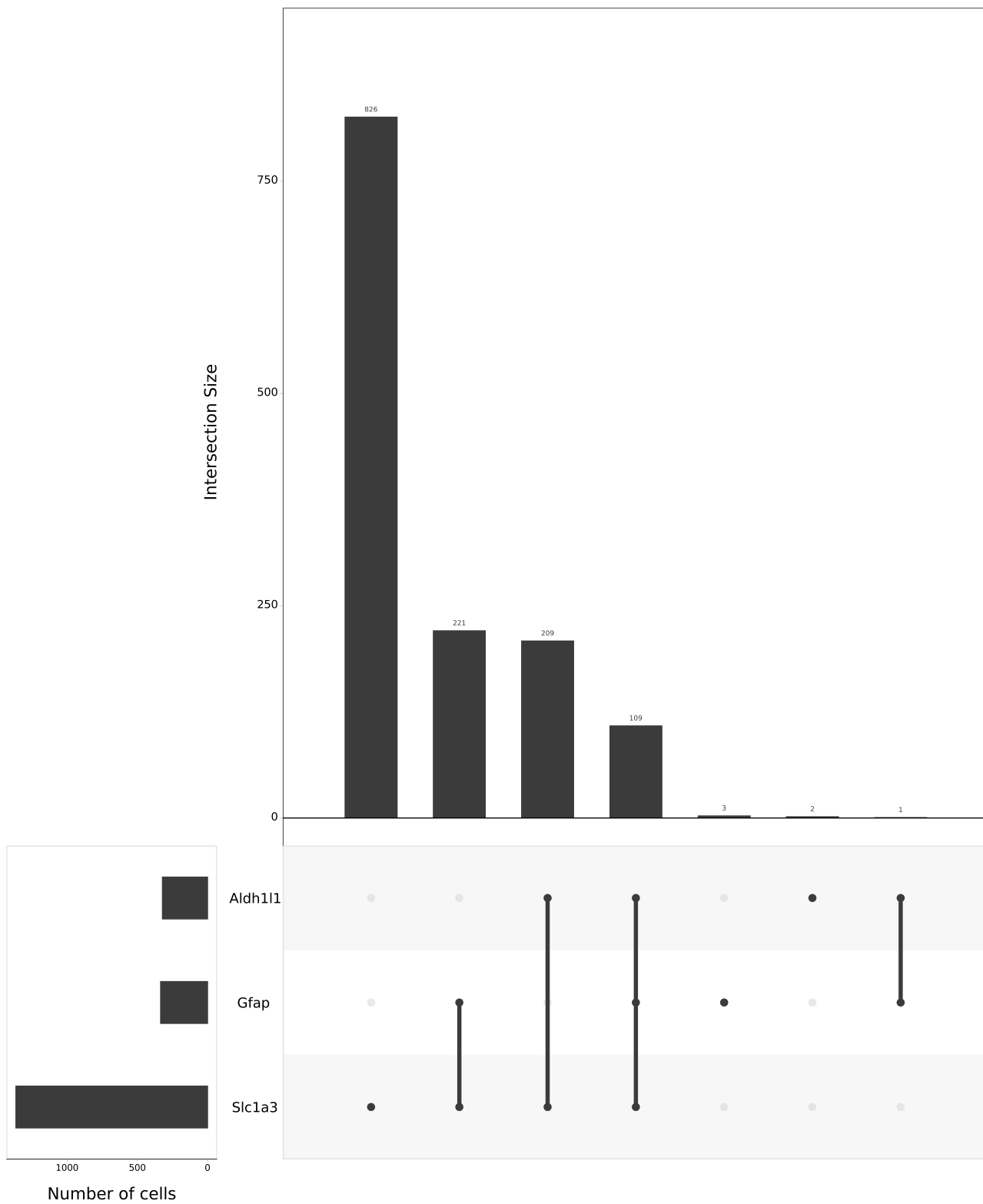
upset(
  as.data.frame(
    content_sbs_mtx |>
      filter(condit == 0) |>
      select(

```

```

        c("Cnr1", "Cnr2", "Gpr55", "Slc1a3", "Gfap", "Aldh111") %>% .[, %in% colnames
    )
),
order.by = "freq",
cutoff = 3,
sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c("Cnr1", "Cnr2", "Gpr55", "Slc1a3", "Gfap", "Aldh111") %>%
  .[, %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

```



```
skim(as.data.frame(
  content_sbs_mtx |>
  filter(condit == 0) |>
  select(
    c("Cnr1", "Cnr2", "Gpr55", "Slc1a3", "Gfap", "Aldh1l1") %>% .[, %in% colname
  )
))
```

Table 25: Data summary

Name	as.data.frame(...)
Number of rows	1392
Number of columns	3
Column type frequency:	
numeric	3
Group variables	None

Variable type: numeric

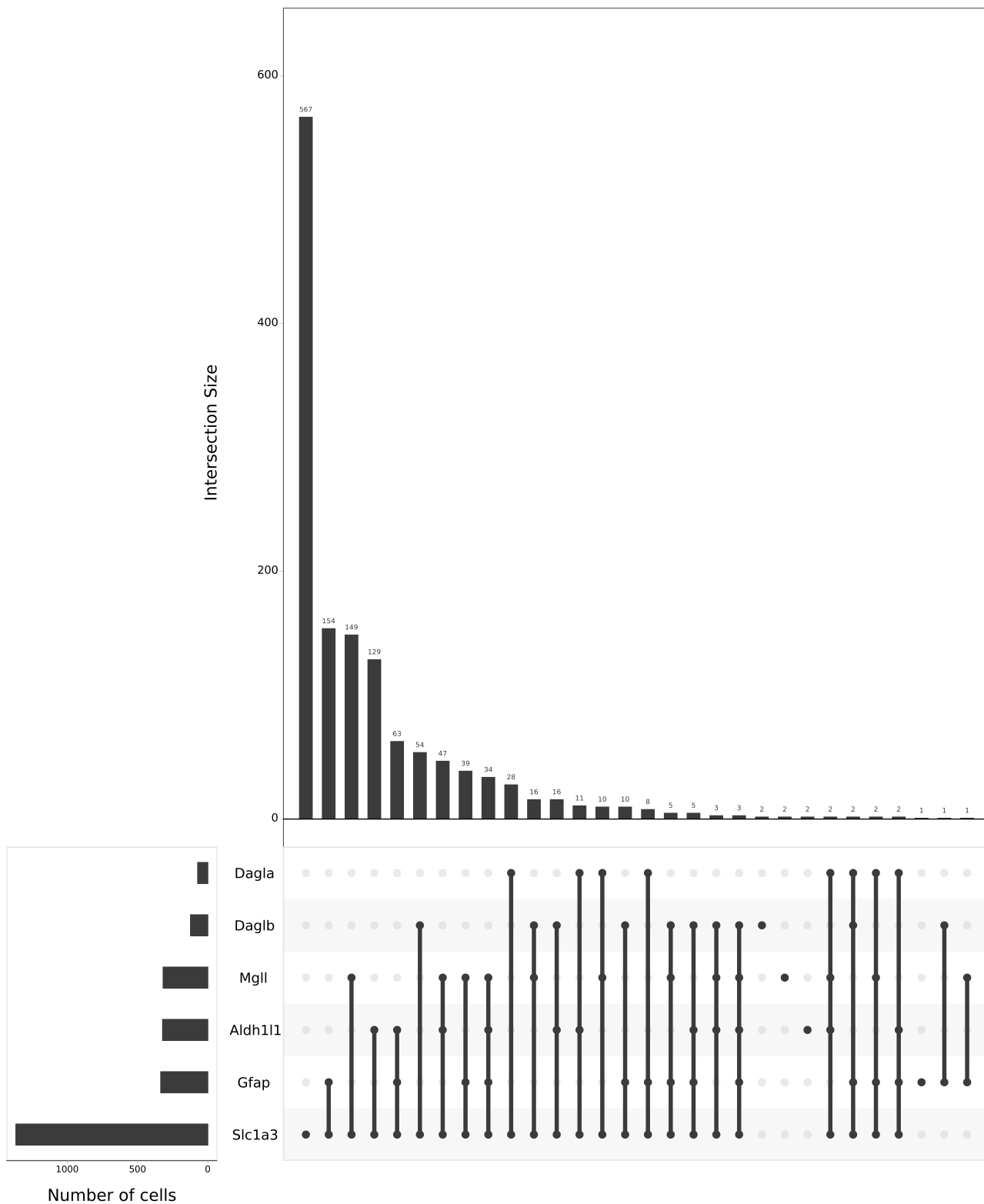
skim_	variablen_	missing	complete_	ratemean	sd	p0	p25	p50	p75	p100	hist
Slc1a3		0	1	0.98	0.14	0	1	1	1	1	
Gfap		0	1	0.24	0.43	0	0	0	0	1	
Aldh1l1		0	1	0.23	0.42	0	0	0	0	1	

```
upset(
  as.data.frame(
    content_sbs_mtx |>
    filter(condit == 0) |>
    select(
      c("Dagla", "Daglb", "Mg1l", "Slc1a3", "Gfap", "Aldh1l1") %>% .[, %in% colname
    )
  ),
  order.by = "freq",
  cutoff = 3,
  sets.x.label = "Number of cells",
  number.angles = 0,
```

```

point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c("Dagla", "Daglb", "Mgll", "Slc1a3", "Gfap", "Aldh1l1") %>%
  [. %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

```



```
skim(as.data.frame(
  content_sbs_mtx |>
  filter(condit == 0) |>
  select(
    c("Dagla", "Daglb", "Mgll", "Slc1a3", "Gfap", "Aldh1l1") %>% .[, %in% colnam
  )
))
```

Table 27: Data summary

Name	as.data.frame(...)
Number of rows	1392
Number of columns	6
Column type frequency:	
numeric	6
Group variables	None

Variable type: numeric

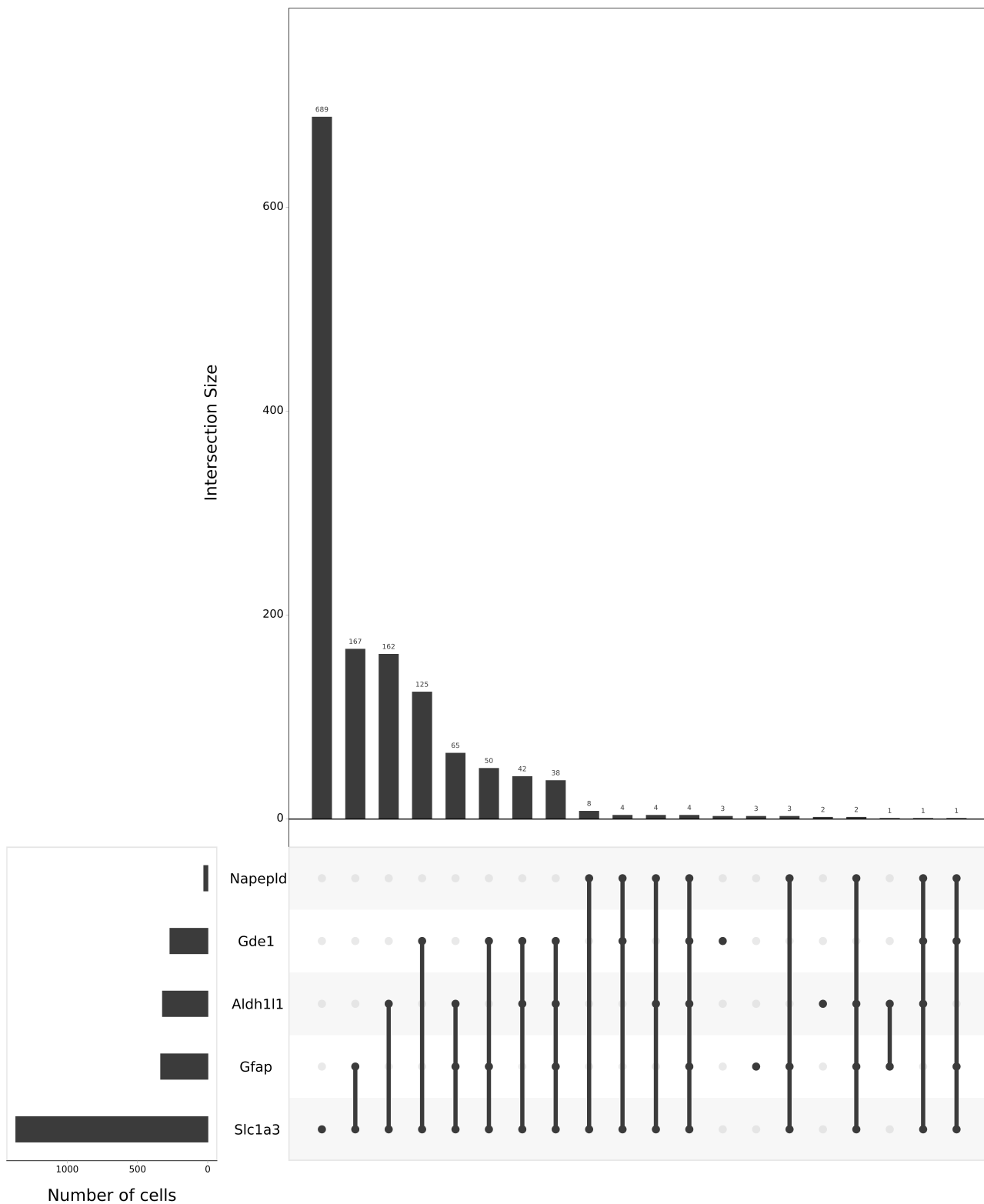
skim_	variablen_	missing	complete_	ratemean	sd	p0	p25	p50	p75	p100	hist
Dagla		0	1	0.05	0.22	0	0	0	0	1	
Daglb		0	1	0.09	0.28	0	0	0	0	1	
Mgll		0	1	0.23	0.42	0	0	0	0	1	
Slc1a3		0	1	0.98	0.14	0	1	1	1	1	
Gfap		0	1	0.24	0.43	0	0	0	0	1	
Aldh1l1		0	1	0.23	0.42	0	0	0	0	1	

```
upset(
  as.data.frame(
    content_sbs_mtx |>
    filter(condit == 0) |>
    select(
      c("Napepld", "Gde1", "Faah", "Slc1a3", "Gfap", "Aldh1l1") %>% .[, %in% colnam
    )
  ),
  order.by = "freq",
```

```

cutoff = 3,
sets.x.label = "Number of cells",
number.angles = 0,
point.size = 3.5, line.size = 2,
text.scale = c(2, 1.6, 2, 1.3, 2, 1.1),
nsets = 30,
nintersects = 30,
sets = c("Napepld", "Gde1", "Faah", "Slc1a3", "Gfap", "Aldh1l1") %>%
  [. %in% colnames(content_sbs_mtx)],
empty.intersections = NULL
)

```

```
skim(as.data.frame(
  content_sbs_mtx |>
    filter(condit == 0) |>
    select(
      c("Napepld", "Gde1", "Faah", "Slc1a3", "Gfap", "Aldh1l1") %>% .[, %in% colna
    )
  ))
```

Table 29: Data summary

Name	as.data.frame(...)
Number of rows	1392
Number of columns	5
Column type frequency:	
numeric	5
Group variables	None

Variable type: numeric

skim_variabl	n_missing	complete	rate	mean	sd	p0	p25	p50	p75	p100	hist
Napepld	0	1	0.02	0.14	0	0	0	0	0	1	
Gde1	0	1	0.19	0.39	0	0	0	0	0	1	
Slc1a3	0	1	0.98	0.14	0	1	1	1	1	1	
Gfap	0	1	0.24	0.43	0	0	0	0	0	1	
Aldh1l1	0	1	0.23	0.42	0	0	0	0	0	1	

```
sessioninfo::session_info()
```

```
- Session info -----
setting  value
version  R version 4.4.0 (2024-04-24)
os       Ubuntu 22.04.4 LTS
system   x86_64, linux-gnu
ui       X11
language en_US:en
```

```

collate en_US.UTF-8
ctype   en_US.UTF-8
tz       Etc/UTC
date     2024-11-25
pandoc   3.2 @ /opt/python/3.8.8/bin/ (via rmarkdown)

```

```

- Packages -----
package      * version      date (UTC) lib source
abind         1.4-8         2024-09-12 [2] RSPM (R 4.4.0)
anndata       * 0.7.5.6      2023-03-17 [2] RSPM (R 4.4.0)
assertthat    0.2.1         2019-03-21 [2] RSPM (R 4.4.0)
base64enc     0.1-3         2015-07-28 [2] RSPM (R 4.4.0)
bayestestR    0.13.2        2024-02-12 [2] RSPM (R 4.4.0)
beeswarm      0.4.0         2021-06-01 [2] RSPM (R 4.4.0)
BiocManager   1.30.23       2024-05-04 [2] RSPM (R 4.4.0)
bit           4.0.5         2022-11-15 [2] RSPM (R 4.4.0)
bit64         4.0.5         2020-08-30 [2] RSPM (R 4.4.0)
circlize      0.4.16        2024-06-19 [2] Github (jokergoo/circlize@9b21578)
cli           3.6.3         2024-06-21 [2] RSPM (R 4.4.0)
cluster       2.1.6         2023-12-01 [2] RSPM (R 4.4.0)
coda          0.19-4.1      2024-01-31 [2] RSPM (R 4.4.0)
codetools     0.2-20        2024-03-31 [2] RSPM (R 4.4.0)
colorspace    2.1-1         2024-07-26 [2] RSPM (R 4.4.0)
correlation   0.8.5         2024-06-16 [2] RSPM (R 4.4.0)
cowplot       * 1.1.3         2024-01-22 [2] RSPM (R 4.4.0)
crayon        1.5.3         2024-06-20 [2] RSPM (R 4.4.0)
data.table    1.16.0        2024-08-27 [2] RSPM (R 4.4.0)
datawizard    0.11.0        2024-06-05 [2] RSPM (R 4.4.0)
deldir        2.0-4         2024-02-28 [2] RSPM (R 4.4.0)
digest        0.6.37        2024-08-19 [2] RSPM (R 4.4.0)
dotCall64     1.1-1         2023-11-28 [2] RSPM (R 4.4.0)
dplyr         * 1.1.4         2023-11-17 [2] RSPM (R 4.4.0)
effectsize    0.8.8         2024-05-12 [2] RSPM (R 4.4.0)
emmeans       1.10.2        2024-05-20 [2] RSPM (R 4.4.0)
estimability  1.5.1         2024-05-12 [2] RSPM (R 4.4.0)
evaluate      1.0.0         2024-09-17 [2] RSPM (R 4.4.0)
fansi         1.0.6         2023-12-08 [2] RSPM (R 4.4.0)
farver        2.1.2         2024-05-13 [2] RSPM (R 4.4.0)
fastDummies   1.7.4         2024-08-16 [2] RSPM (R 4.4.0)
fastmap       1.2.0         2024-05-15 [2] RSPM (R 4.4.0)
fitdistrplus  1.2-1         2024-07-12 [2] RSPM (R 4.4.0)

```

forcats	* 1.0.0	2023-01-29	[2]	RSPM (R 4.4.0)
future	* 1.34.0	2024-07-29	[2]	RSPM (R 4.4.0)
future.apply	1.11.2	2024-03-28	[2]	RSPM (R 4.4.0)
generics	0.1.3	2022-07-05	[2]	RSPM (R 4.4.0)
ggbeeswarm	0.7.2	2024-06-19	[2]	Github (eclarke/ggbeeswarm@ce2da8a)
ggmin	0.0.0.9000	2024-06-19	[2]	Github (sjessa/ggmin@8ada274)
ggplot2	* 3.5.1	2024-04-23	[2]	RSPM (R 4.4.0)
ggprism	1.0.5	2024-06-19	[2]	Github (csdaw/ggprism@b6e6c0e)
ggrastr	1.0.2	2024-06-19	[2]	Github (VPetukhov/ggrastr@50ca3e0)
ggrepel	0.9.6	2024-09-20	[2]	Github (slowkow/ggrepel@e94776b)
ggribes	0.5.6	2024-01-23	[2]	RSPM (R 4.4.0)
ggstatsplot	* 0.12.3.9000	2024-06-19	[2]	Github (IndrajeetPatil/ggstatsplot@d55f
GlobalOptions	0.1.2	2020-06-10	[2]	RSPM (R 4.4.0)
globals	0.16.3	2024-03-08	[2]	RSPM (R 4.4.0)
glue	1.7.0	2024-01-09	[2]	RSPM (R 4.4.0)
goftest	1.2-3	2021-10-07	[2]	RSPM (R 4.4.0)
gridExtra	2.3	2017-09-09	[2]	RSPM (R 4.4.0)
gtable	0.3.5	2024-04-22	[2]	RSPM (R 4.4.0)
hdf5r	1.3.10	2024-03-02	[2]	RSPM (R 4.4.0)
here	* 1.0.1	2020-12-13	[2]	RSPM (R 4.4.0)
hms	1.1.3	2023-03-21	[2]	RSPM (R 4.4.0)
htmltools	0.5.8.1	2024-04-04	[2]	RSPM (R 4.4.0)
htmlwidgets	1.6.4	2023-12-06	[2]	RSPM (R 4.4.0)
httpuv	1.6.15	2024-03-26	[2]	RSPM (R 4.4.0)
httr	1.4.7	2023-08-15	[2]	RSPM (R 4.4.0)
ica	1.0-3	2022-07-08	[2]	RSPM (R 4.4.0)
igraph	2.0.3	2024-03-13	[2]	RSPM (R 4.4.0)
insight	0.20.1	2024-06-11	[2]	RSPM (R 4.4.0)
irlba	2.3.5.1	2022-10-03	[2]	RSPM (R 4.4.0)
janitor	2.2.0.9000	2024-06-19	[2]	Github (sfirke/janitor@80cd1eb)
jsonlite	1.8.8	2023-12-04	[2]	RSPM (R 4.4.0)
KernSmooth	2.23-24	2024-05-17	[2]	RSPM (R 4.4.0)
knitr	1.48	2024-07-07	[2]	RSPM (R 4.4.0)
labeling	0.4.3	2023-08-29	[2]	RSPM (R 4.4.0)
later	1.3.2	2023-12-06	[2]	RSPM (R 4.4.0)
lattice	0.22-6	2024-03-20	[2]	RSPM (R 4.4.0)
lazyeval	0.2.2	2019-03-15	[2]	RSPM (R 4.4.0)
leiden	0.4.3.1	2023-11-17	[2]	RSPM (R 4.4.0)
lifecycle	1.0.4	2023-11-07	[2]	RSPM (R 4.4.0)
listenv	0.9.1	2024-01-29	[2]	RSPM (R 4.4.0)
lmtest	0.9-40	2022-03-21	[2]	RSPM (R 4.4.0)

lubridate	* 1.9.3	2023-09-27	[2]	RSPM	(R 4.4.0)
magrittr	* 2.0.3	2022-03-30	[2]	RSPM	(R 4.4.0)
MASS	7.3-61	2024-06-13	[2]	RSPM	(R 4.4.0)
Matrix	1.7-0	2024-04-26	[2]	RSPM	(R 4.4.0)
matrixStats	1.4.1	2024-09-08	[2]	RSPM	(R 4.4.0)
mime	0.12	2021-09-28	[2]	RSPM	(R 4.4.0)
miniUI	0.1.1.1	2018-05-18	[2]	RSPM	(R 4.4.0)
multcomp	1.4-25	2023-06-20	[2]	RSPM	(R 4.4.0)
munsell	0.5.1	2024-04-01	[2]	RSPM	(R 4.4.0)
mvtnorm	1.2-5	2024-05-21	[2]	RSPM	(R 4.4.0)
nlme	3.1-165	2024-06-06	[2]	RSPM	(R 4.4.0)
paletteer	1.6.0	2024-01-21	[2]	RSPM	(R 4.4.0)
parallelly	1.38.0	2024-07-27	[2]	RSPM	(R 4.4.0)
parameters	0.21.7	2024-05-14	[2]	RSPM	(R 4.4.0)
patchwork	* 1.3.0.9000	2024-09-20	[2]	Github	(thomasp85/patchwork@2695a9f)
pbapply	1.7-2	2023-06-27	[2]	RSPM	(R 4.4.0)
pillar	1.9.0	2023-03-22	[2]	RSPM	(R 4.4.0)
pkgconfig	2.0.3	2019-09-22	[2]	RSPM	(R 4.4.0)
plotly	4.10.4	2024-01-13	[2]	RSPM	(R 4.4.0)
plyr	1.8.9	2023-10-02	[2]	RSPM	(R 4.4.0)
png	0.1-8	2022-11-29	[2]	RSPM	(R 4.4.0)
polyclip	1.10-7	2024-07-23	[2]	RSPM	(R 4.4.0)
progressr	0.14.0	2023-08-10	[2]	RSPM	(R 4.4.0)
promises	1.3.0	2024-04-05	[2]	RSPM	(R 4.4.0)
purrr	* 1.0.2	2023-08-10	[2]	RSPM	(R 4.4.0)
R.methodsS3	1.8.2	2022-06-13	[2]	RSPM	(R 4.4.0)
R.oo	1.26.0	2024-01-24	[2]	RSPM	(R 4.4.0)
R.utils	2.12.3	2023-11-18	[2]	RSPM	(R 4.4.0)
R6	2.5.1	2021-08-19	[2]	RSPM	(R 4.4.0)
RANN	2.6.2	2024-08-25	[2]	RSPM	(R 4.4.0)
RColorBrewer	* 1.1-3	2022-04-03	[2]	RSPM	(R 4.4.0)
Rcpp	1.0.13	2024-07-17	[2]	RSPM	(R 4.4.0)
RcppAnnoy	0.0.22	2024-01-23	[2]	RSPM	(R 4.4.0)
RcppHNSW	0.6.0	2024-02-04	[2]	RSPM	(R 4.4.0)
readr	* 2.1.5	2024-01-10	[2]	RSPM	(R 4.4.0)
rematch2	2.1.2	2020-05-01	[2]	RSPM	(R 4.4.0)
remotes	2.5.0	2024-03-17	[2]	RSPM	(R 4.4.0)
repr	1.1.7	2024-03-22	[2]	RSPM	(R 4.4.0)
reshape2	1.4.4	2020-04-09	[2]	RSPM	(R 4.4.0)
reticulate	* 1.39.0	2024-09-05	[2]	RSPM	(R 4.4.0)
rlang	1.1.4	2024-06-04	[2]	RSPM	(R 4.4.0)

rmarkdown	2.28	2024-08-17	[2]	RSPM (R 4.4.0)
ROCR	1.0-11	2020-05-02	[2]	RSPM (R 4.4.0)
rprojroot	2.0.4	2023-11-05	[2]	RSPM (R 4.4.0)
RSpectra	0.16-2	2024-07-18	[2]	RSPM (R 4.4.0)
rstudioapi	0.16.0	2024-03-24	[2]	RSPM (R 4.4.0)
rsvd	1.0.5	2021-04-16	[2]	RSPM (R 4.4.0)
Rtsne	0.17	2023-12-07	[2]	RSPM (R 4.4.0)
sandwich	3.1-0	2023-12-11	[2]	RSPM (R 4.4.0)
scales	1.3.0	2023-11-28	[2]	RSPM (R 4.4.0)
scattermore	1.2	2023-06-12	[2]	RSPM (R 4.4.0)
scCustomize	* 2.1.2	2024-06-19	[2]	Github (samuel-marsh/scCustomize@fc7a28
sceasy	* 0.0.7	2024-06-19	[2]	Github (cellgeni/sceasy@c1c0bf9)
sctransform	0.4.1	2023-10-19	[2]	RSPM (R 4.4.0)
sessioninfo	1.2.2	2021-12-06	[2]	RSPM (R 4.4.0)
Seurat	* 5.1.0.9005	2024-09-20	[2]	Github (satijalab/seurat@95de9dc)
SeuratDisk	* 0.0.0.9021	2024-06-19	[2]	Github (mojaveazure/seurat-disk@877d4e1
SeuratObject	* 5.0.99.9001	2024-09-20	[2]	Github (satijalab/seurat-object@1a140c7
SeuratWrappers	* 0.3.5	2024-06-19	[2]	Github (satijalab/seurat-wrappers@8d46d
shape	1.4.6.1	2024-02-23	[2]	RSPM (R 4.4.0)
shiny	1.9.1	2024-08-01	[2]	RSPM (R 4.4.0)
skimr	* 2.1.5	2024-06-19	[2]	Github (ropensci/skimr@d5126aa)
snakecase	0.11.1	2023-08-27	[2]	RSPM (R 4.4.0)
sp	* 2.1-4	2024-04-30	[2]	RSPM (R 4.4.0)
spam	2.10-0	2023-10-23	[2]	RSPM (R 4.4.0)
spatstat.data	3.1-2	2024-06-21	[2]	RSPM (R 4.4.0)
spatstat.explore	3.3-2	2024-08-21	[2]	RSPM (R 4.4.0)
spatstat.geom	3.3-3	2024-09-18	[2]	RSPM (R 4.4.0)
spatstat.random	3.3-2	2024-09-18	[2]	RSPM (R 4.4.0)
spatstat.sparse	3.1-0	2024-06-21	[2]	RSPM (R 4.4.0)
spatstat.univar	3.0-1	2024-09-05	[2]	RSPM (R 4.4.0)
spatstat.utils	3.1-0	2024-08-17	[2]	RSPM (R 4.4.0)
statsExpressions	1.5.4	2024-03-20	[2]	RSPM (R 4.4.0)
stringi	1.8.4	2024-05-06	[2]	RSPM (R 4.4.0)
stringr	* 1.5.1	2023-11-14	[2]	RSPM (R 4.4.0)
survival	3.7-0	2024-06-05	[2]	RSPM (R 4.4.0)
tensor	1.5	2012-05-05	[2]	RSPM (R 4.4.0)
TH.data	1.1-2	2023-04-17	[2]	RSPM (R 4.4.0)
tibble	* 3.2.1	2023-03-20	[2]	RSPM (R 4.4.0)
tidyr	* 1.3.1	2024-01-24	[2]	RSPM (R 4.4.0)
tidyselect	1.2.1	2024-03-11	[2]	RSPM (R 4.4.0)
tidyverse	* 2.0.0.9000	2024-06-19	[2]	Github (tidyverse/tidyverse@62f32d4)

timechange	0.3.0	2024-01-18	[2]	RSPM (R 4.4.0)
tzdb	0.4.0	2023-05-12	[2]	RSPM (R 4.4.0)
UpSetR	* 1.4.0	2024-06-19	[2]	Github (hms-dbmi/UpSetR@b14854a)
utf8	1.2.4	2023-10-22	[2]	RSPM (R 4.4.0)
uwot	0.2.2	2024-04-21	[2]	RSPM (R 4.4.0)
vctrs	0.6.5	2023-12-01	[2]	RSPM (R 4.4.0)
vipor	0.4.7	2023-12-18	[2]	RSPM (R 4.4.0)
viridis	* 0.6.5	2024-01-29	[2]	RSPM (R 4.4.0)
viridisLite	* 0.4.2	2023-05-02	[2]	RSPM (R 4.4.0)
vroom	1.6.5	2023-12-05	[2]	RSPM (R 4.4.0)
withr	3.0.1	2024-07-31	[2]	RSPM (R 4.4.0)
xfun	0.47	2024-08-17	[2]	RSPM (R 4.4.0)
xtable	1.8-4	2019-04-21	[2]	RSPM (R 4.4.0)
yaml	2.3.10	2024-07-26	[2]	RSPM (R 4.4.0)
zeallot	0.1.0	2018-01-28	[2]	RSPM (R 4.4.0)
zoo	1.8-12	2023-04-13	[2]	RSPM (R 4.4.0)

[1] /home/etretiakov/R/x86_64-pc-linux-gnu-library/4.4

[2] /opt/R/4.4.0/lib/R/library

- Python configuration -----

python:	/opt/python/3.8.8/bin/python
libpython:	/opt/python/3.8.8/lib/libpython3.8.so
pythonhome:	/opt/python/3.8.8:/opt/python/3.8.8
version:	3.8.8 packaged by conda-forge (default, Feb 20 2021, 16:22:27) [
numpy:	/opt/python/3.8.8/lib/python3.8/site-packages/numpy
numpy_version:	1.23.5
scanpy:	/home/etretiakov/.local/lib/python3.8/site-packages/scanpy

NOTE: Python version was forced by RETICULATE_PYTHON
