

Adreanal Medulla Reclustering

```
library(Seurat)
library(ggplot2)
```

Loading required package: SeuratObject

Loading required package: sp

Attaching package: 'SeuratObject'

The following objects are masked from 'package:base':

intersect, t

Load the data

```
seu <- readRDS("data/processed/integrated_annotated.rds")
```

Subset AMC (Schwann cell precursors (SCPs), Chromaffin cells, Sympathoblasts)

```
unique(seu$celltype)
```

1. Adrenal gland cortex
2. Chromaffin cells
3. Endothelium
4. 10
5. Subepicardial and abdominal mesenchyme
6. Intermediate mesoderm
7. Sympathoblasts
8. Erythroid cells
9. HSCs
10. 9
11. Immune cells
12. SCPs

Levels: 1. 'SCPs' 2. 'Chromaffin cells' 3. 'Sympathoblasts' 4. 'Adrenal gland cortex' 5. 'Subepicardial and abdominal mesenchyme' 6. 'Endothelium' 7. 'Intermediate mesoderm' 8. 'HSCs' 9. 'Immune cells' 10. 'Erythroid cells' 11. '9' 12. '10'

```
seu_subset <- subset(seu, subset = celltype %in% c("SCPs", "Chromaffin cells", "Sympathoblasts"))
```

```
unique(seu_subset$celltype)
```

1. Chromaffin cells
2. Sympathoblasts
3. SCPs

Levels: 1. 'SCPs' 2. 'Chromaffin cells' 3. 'Sympathoblasts' 4. 'Adrenal gland cortex' 5. 'Subepicardial and abdominal mesenchyme' 6. 'Endothelium' 7. 'Intermediate mesoderm' 8. 'HSCs' 9. 'Immune cells' 10. 'Erythroid cells' 11. '9' 12. '10'

Recluster

```
seu_subset <- FindVariableFeatures(seu_subset)
```

Warning message in FindVariableFeatures.Assay(object = object[[assay]], selection.method = selection.method, selection.method set to 'vst' but count slot is empty; will use data slot instead"

Warning message in eval(predvars, data, env):

"NaNs produced"

Warning message in hvf.info\$variance.expected[not.const] <- 10~fit\$fitted:

"number of items to replace is not a multiple of replacement length"

```
seu_subset <- ScaleData(seu_subset)
seu_subset <- RunPCA(seu_subset)
seu_subset <- FindNeighbors(seu_subset, dims = 1:30)
```

Centering and scaling data matrix

PC_ 1

Positive: STMN2, RGS5, CHGB, DBH, PCSK1N, EEF1A2, HAND2-AS1, EML5, CD24, BEX1
GATA3, HAND2, SYT1, MIAT, MAP1B, GATA2, CHGA, CNTN1, BASP1, PHOX2A
TUBB2B, ELAVL4, ISL1, CHRNA3, DPP6, SEZ6L2, RAMP1, RGS4, KIF21A, ELAVL3

Negative: PLP1, PTPRZ1, EDNRB, COL5A2, ERBB3, OLFML2A, MPZ, SPARC, S100B, CDH19
VCAN, TGFBR2, POSTN, MOXD1, TRPM3, NR2F2, ABCA8, GPM6B, TTYH1, PLAT
METRN, NID1, HSPG2, COL2A1, SOX10, LMO4, COL1A1, LGI4, GAS7, PLEKHA4

PC_ 2

Positive: TUBB, SOX4, STMN1, TUBA1A, RTN1, GAP43, TUBB2B, TMSB15A, STMN4, ELAVL4
SOX11, TUBA1B, SLC6A2, HMGB1, SYNE2, SMC4, TMSB4X, ASPM, SLC38A1, MLLT11
KIF21A, PLXNA4, RBFOX1, VIM, PHOX2B, TOP2A, PRC1, MKI67, JPT1, TMPO

Negative: DLK1, PENK, SLC24A2, ST18, INSM1, CCSER1, PNMT, VWA5B2, NDUFA4L2, CHGA
ADM, ARC, SERTM2, C1QL1, CDKN1C, SCARB1, F10, TH, CCND2, HTATSF1
PCSK2, VEGFA, JUNB, ROBO2, DGKK, SCG2, CARTPT, HIST1H2AC, SMIM1, RALYL

PC_ 3

Positive: CENPF, MKI67, TOP2A, NUSAP1, CENPE, CDK1, ASPM, NUF2, GTSE1, TTK

RRM2, UBE2C, KIF23, CCNB1, TPX2, PBK, CCNA2, PIMREG, HMGB2, CDCA8
 NCAPG, AURKB, PTTG1, CDKN3, PLK1, FOXM1, DLGAP5, BUB1, KIF11, BIRC5
 Negative: SOX4, PRPH, STMN4, MAOA, TUBA1A, GAP43, CCND1, GAL, TUBB2B, NEFL
 RBFOX1, RTN1, NPY, C4orf48, PLPPR3, PLXNA4, UCHL1, MAPT, DPYSL3, PHOX2B
 ARHGDIG, SCN9A, SCN3B, TUBB2A, ANXA2, TBX2, STMN2, MAP1B, ANK2, HBG2

PC_4

Positive: TMEM88, PLVAP, CD93, CD34, EMCN, F8, ROBO4, TIE1, FLT1, SOX18
 CD109, RASGRP3, CAVIN2, A2M, CDH5, ADGRF5, TFPI, CALCRL, VAMP5, BTNL9
 LMO2, HLA-E, TM4SF18, KDR, NOSTRIN, PECAM1, LRRC32, ICAM2, LYVE1, PCDH12
 Negative: NUF2, MKI67, CENPF, KIFC1, GAS2L3, ASPM, PTTG1, DST, SLITRK2, TROAP
 DLGAP5, MXD3, KIF2C, FLRT3, KIF4A, UBE2C, RTKN2, ZEB2, AURKB, MNS1
 EGFLAM, FOXM1, NCAPG, PLP1, TPX2, BIRC5, GTSE1, COL25A1, ERBB3, S100B

PC_5

Positive: NR5A1, MGARP, CYP11A1, FDXR, ASB4, SIGLEC11, STAR, APOE, TCEA3, ALDH1A2
 GIPC2, MRAP, SNCG, CCDC141, MT3, MC2R, NRK, MCF2, ZG16B, AC020571.1
 FDX1, DHCR24, SULT2A1, GRB14, RBM47, NOV, TNNI3, INHA, CYP17A1, MAP3K15
 Negative: PLVAP, FLT1, CALCRL, CAVIN2, F8, PECAM1, CD93, TM4SF18, CLDN5, BTNL9
 CDH5, CD34, ICAM2, SOX18, TMEM88, TIE1, TEK, ROBO4, PROCR, KDR
 EMCN, ADGRF5, PCAT19, FGF23, CETP, RASGRP3, HLA-E, IRX3, EHD3, CEACAM1

Computing nearest neighbor graph

Computing SNN

```
seu_subset <- FindClusters(seu_subset, resolution = 0.1)
seu_subset <- RunUMAP(seu_subset, dims = 1:40)
```

Modularity Optimizer version 1.3.0 by Ludo Waltman and Nees Jan van Eck

Number of nodes: 2118

Number of edges: 94548

Running Louvain algorithm...

Maximum modularity in 10 random starts: 0.9551

Number of communities: 4

Elapsed time: 0 seconds

20:32:52 UMAP embedding parameters a = 0.9922 b = 1.112

20:32:52 Read 2118 rows and found 40 numeric columns

20:32:52 Using Annoy for neighbor search, n_neighbors = 30

20:32:52 Building Annoy index with metric = cosine, n_trees = 50

0% 10 20 30 40 50 60 70 80 90 100%

[----|----|----|----|----|----|----|----|----|----|

*
 *

```
20:32:53 Writing NN index file to temp file /var/folders/wl/jrkngsm57b944tj7rtjg12000000gn/T//Rtmpal
```

```

20:32:53 Searching Annoy index using 1 thread, search_k = 3000

20:32:53 Annoy recall = 100%

20:32:53 Commencing smooth kNN distance calibration using 1 thread
with target n_neighbors = 30

20:32:53 Initializing from normalized Laplacian + noise (using RSpectra)

20:32:53 Commencing optimization for 500 epochs, with 93720 positive edges

20:32:53 Using rng type: pcg

20:32:54 Optimization finished

```

```
DimPlot(seu_subset)
```

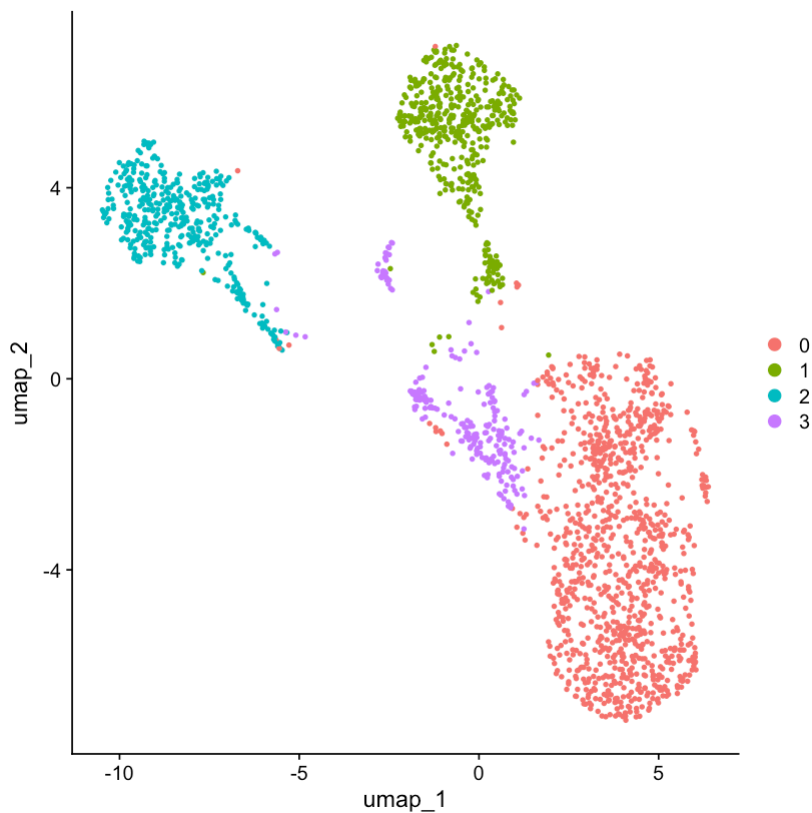


Fig 2 Markers from the paper

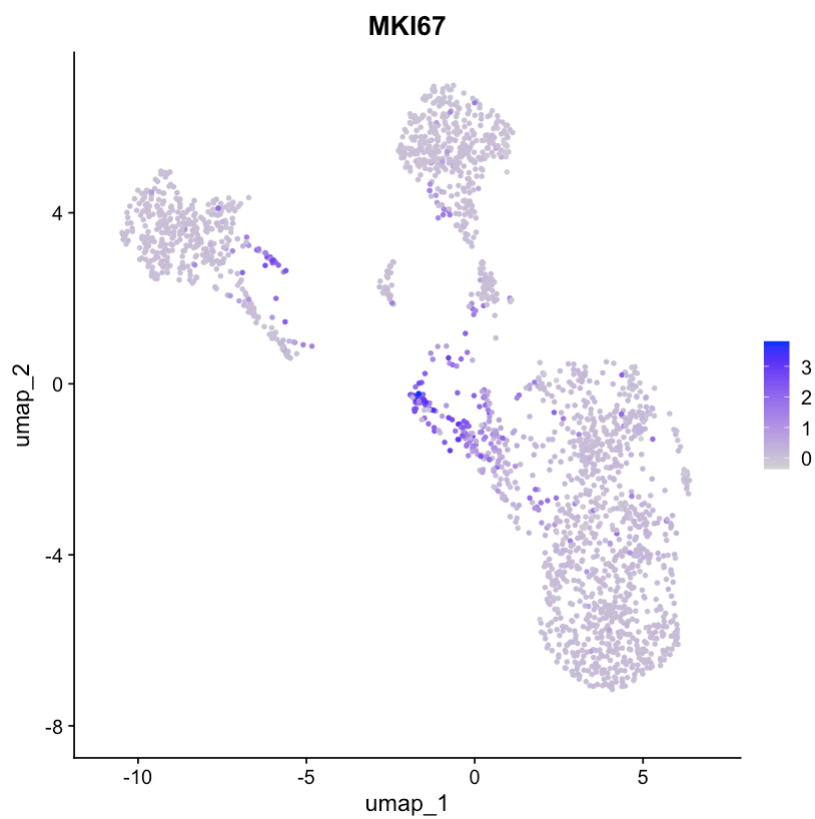
```

markers <- list(
  "Prolifertating sympathoblasts" = "MKI67",
  "Sympathoblasts" = c("ELAVL4", "ISL1", "PRPH"),
  "SCPs" = c("SOX10", "PLP1"),
  "Chromaffin cells" = c("CHGA", "PNMT")
)

```

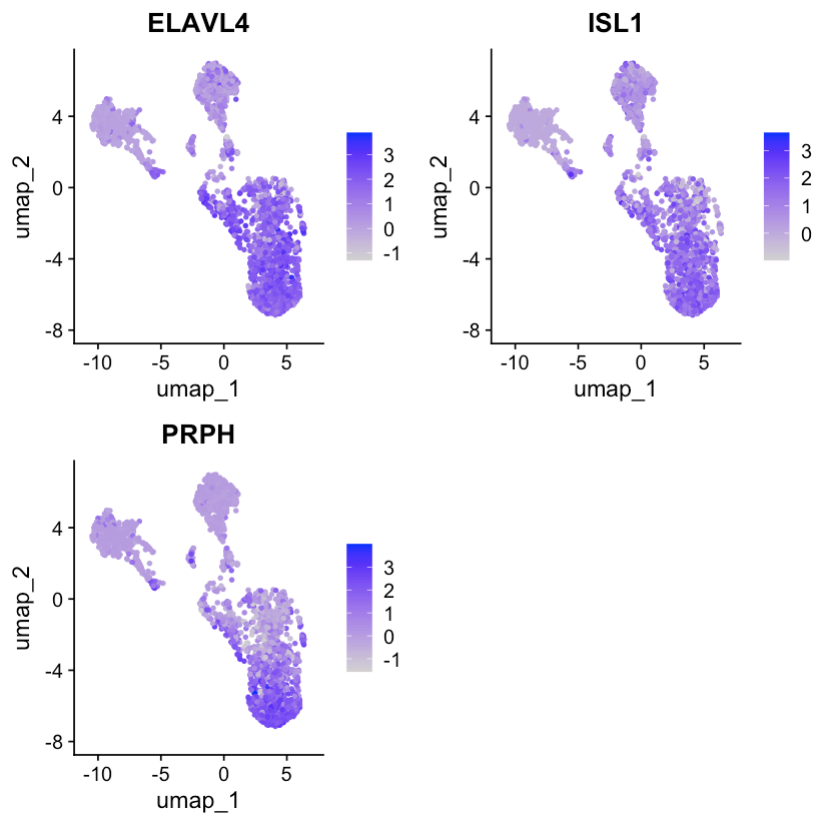
Markers for Prol. Sympathoblasts

```
FeaturePlot(seu_subset,markers[[1]])  
# cluster 3
```



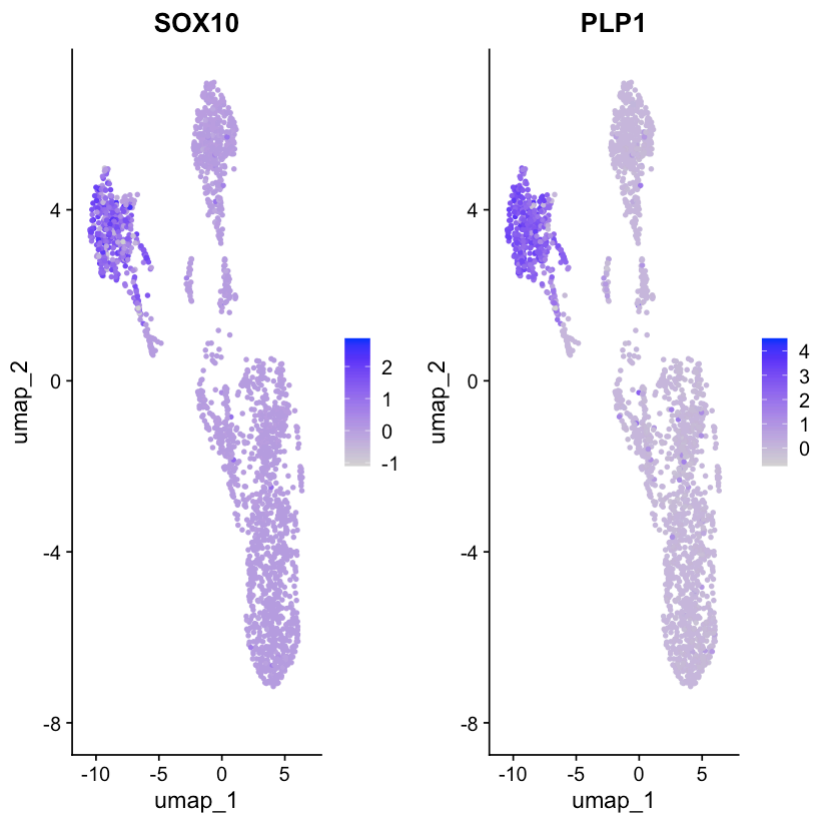
Markers for Sympathoblasts

```
FeaturePlot(seu_subset,markers[[2]])  
# cluster 0
```



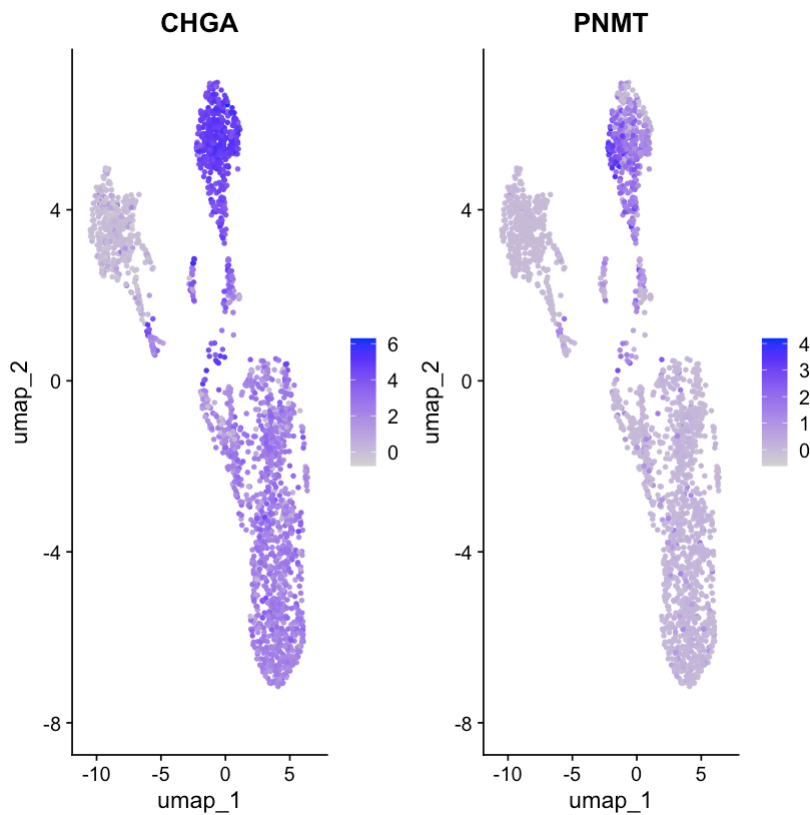
Markers for SCPs

```
FeaturePlot(seu_subset, markers[[3]], ncol=2)  
# cluster 2
```



Chromaffin cells

```
FeaturePlot(seu_subset, markers[[4]], ncol=2)  
# cluster 1
```

```
cluster <- list(
  "3" = "Prolifertating sympathoblasts",# = "MKI67",
  "0" = "Sympathoblasts",# = c("ELAVL4", "ISL1", "PRPH"),
  "2" = "SCPs",# = c("SOX10","PLP1"),
  "1" = "Chromaffin cells"# = c("CHGA","PNMT")
)
```

```
cluster <- unlist(cluster)
seu_subset <- RenameIdents(seu_subset, cluster)
seu_subset$celltype <- Idents(seu_subset)
```

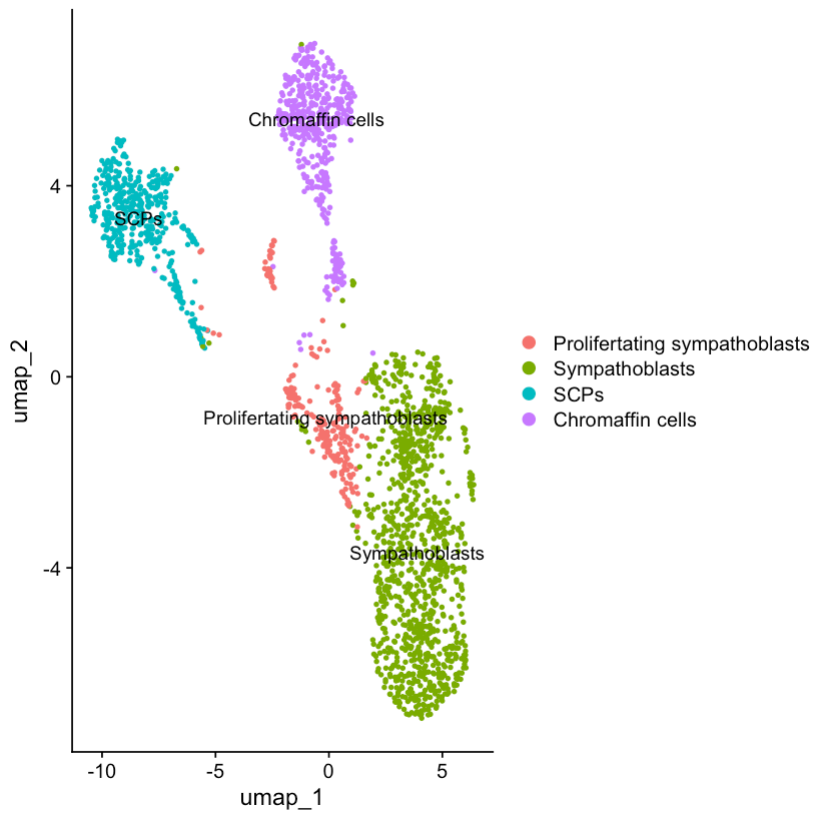
```
unique(seu_subset$celltype)
```

1. Chromaffin cells
2. Sympathoblasts
3. Prolifertating sympathoblasts
4. SCs

Levels: 1. 'Prolifertating sympathoblasts' 2. 'Sympathoblasts' 3. 'SCs' 4. 'Chromaffin cells'

UMAP with Annotations

```
umap_ann_subset =DimPlot(seu_subset, label=TRUE)
umap_ann_subset
```



```
ggsave("plots/AMC_subset_UMAP_annotated.pdf",umap_ann_subset, width = 10, height = 8)
```

```
saveRDS(seu_subset,"data/processed/AMC_subset_annotated.rds")
```