UMAP on murine fetal brain 1k cells

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Loading necessary packages

```
library (dplyr, quietly = TRUE)

##
## 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

library (Seurat)
```

Direct R to the directory you want to work in

setwd ("/Users/nr267/Desktop/All Work/Classes/Spring 2020/Seurat class resources/")

Load data and carry out QC

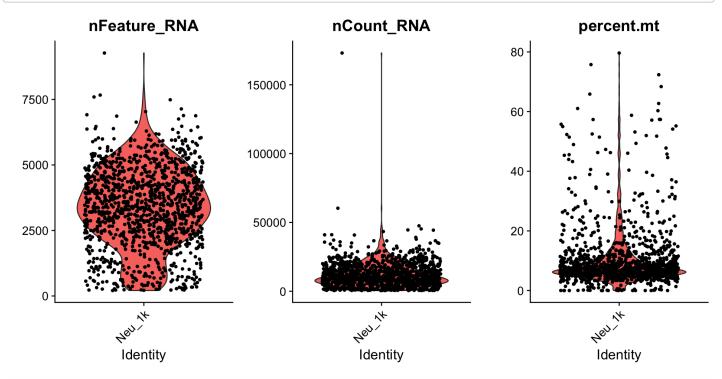
```
Neu_1k <- Read10X(data.dir = "../Seurat_class_resources/filtered_feature_bc_matrix_1K_ne
uron/1K_mouse_brain_cells/")
Neu_1k <- CreateSeuratObject(counts = Neu_1k, project = "Neu_1k", min.cells = 3, min.fea
tures = 200)
Neu_1k</pre>
```

```
## An object of class Seurat
## 15657 features across 1243 samples within 1 assay
## Active assay: RNA (15657 features)
```

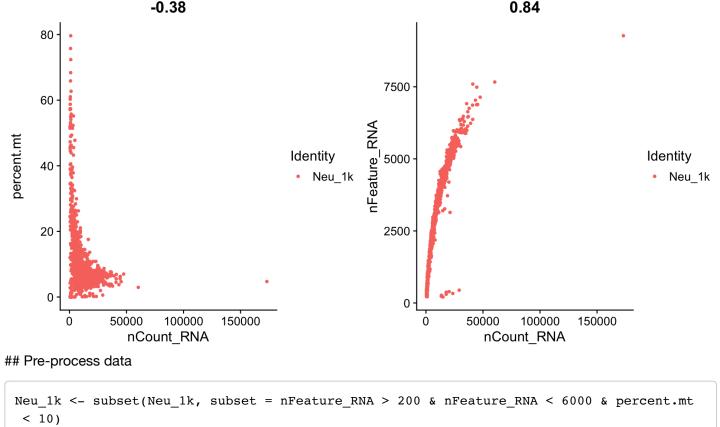
```
Neu_1k[["percent.mt"]] <- PercentageFeatureSet(Neu_1k, pattern = "^mt-")
head (Neu_1k@meta.data, 10)</pre>
```

```
##
                    orig.ident nCount_RNA nFeature_RNA percent.mt
                                      1023
## AAACGAATCAAAGCCT
                        Neu_1k
                                                    417
                                                         52.492669
## AAACGCTGTAATGTGA
                        Neu 1k
                                      7209
                                                   2743
                                                         12.109863
## AAACGCTGTCCTGGGT
                        Neu_1k
                                      9881
                                                   3601
                                                          7.600445
## AAAGAACCAGGACATG
                        Neu 1k
                                      5289
                                                   2385
                                                          5.483078
## AAAGGTACACACGGTC
                        Neu_1k
                                     18674
                                                   4746
                                                          5.515690
## AAAGTCCAGTCACTAC
                        Neu_1k
                                     20189
                                                   5097
                                                          5.735797
## AAAGTCCTCCAGCCTT
                        Neu_1k
                                      6447
                                                   2557 12.936249
## AAAGTGAGTTCCTAAG
                        Neu_1k
                                     16836
                                                   4710
                                                          4.793300
## AAAGTGATCAGTGGGA
                                                   1007
                                                         32.670868
                        Neu_1k
                                      2063
## AAATGGAGTAGCGTCC
                        Neu_1k
                                      1946
                                                    924
                                                         31.397739
```

```
VlnPlot (Neu_1k, features = c("nFeature_RNA", "nCount_RNA", "percent.mt"), ncol = 3)
```



```
plot1 <- FeatureScatter(Neu_1k, feature1 = "nCount_RNA", feature2 = "percent.mt")
plot2 <- FeatureScatter(Neu_1k, feature1 = "nCount_RNA", feature2 = "nFeature_RNA")
CombinePlots(plots = list(plot1, plot2))</pre>
```



Check data dimensions after subsetting

```
Meu_1k

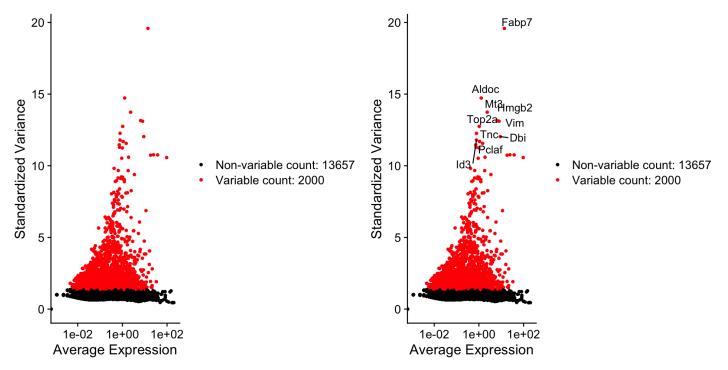
## An object of class Seurat
## 15657 features across 862 samples within 1 assay
## Active assay: RNA (15657 features)
```

LogNormalize Data (normalize count data per cell and transform to log scale)

```
Neu_1k <- NormalizeData(Neu_1k, verbose = TRUE)</pre>
```

Pre-process data for PCA

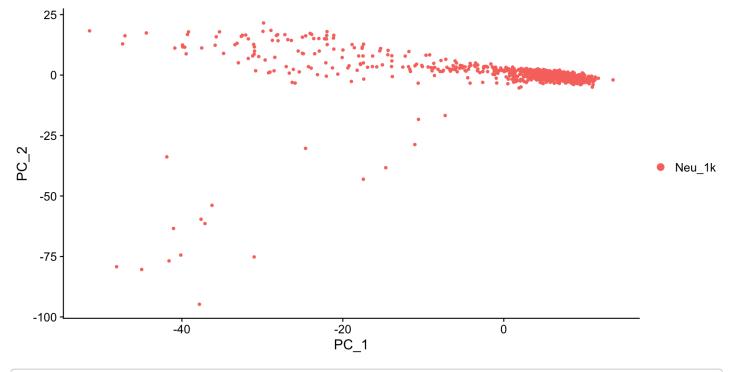
```
Neu_1k <- FindVariableFeatures(Neu_1k, selection.method = "vst", nfeatures = 2000)
top10 <- head(VariableFeatures(Neu_1k), 10)
plot3 <- VariableFeaturePlot(Neu_1k)
plot4 <- LabelPoints(plot = plot3, points = top10, repel = TRUE)
CombinePlots(plots = list(plot3, plot4))</pre>
```



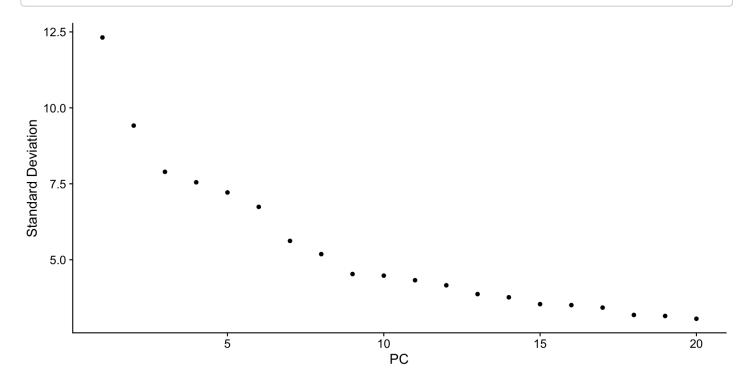
```
all.genes <- rownames(Neu_1k)
Neu_1k <- ScaleData(Neu_1k, features = all.genes)
Neu_1k</pre>
```

```
## An object of class Seurat
## 15657 features across 862 samples within 1 assay
## Active assay: RNA (15657 features)
```

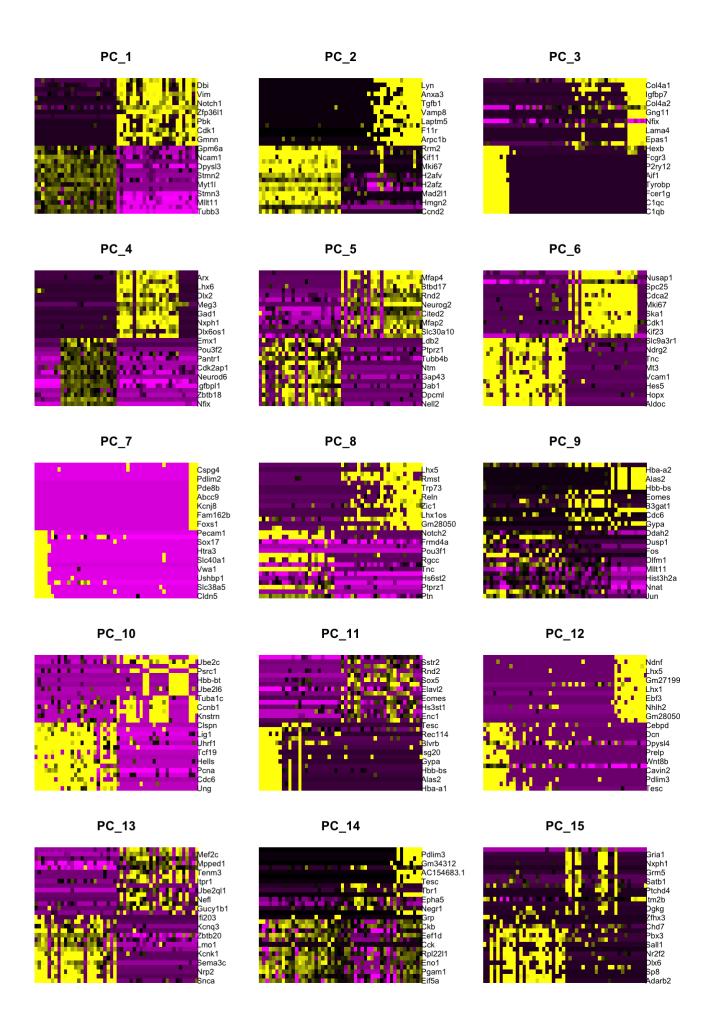
```
Neu_1k<- RunPCA(Neu_1k, features = VariableFeatures(object = Neu_1k))
DimPlot (Neu_1k, reduction = "pca")</pre>
```



ElbowPlot(Neu_1k)



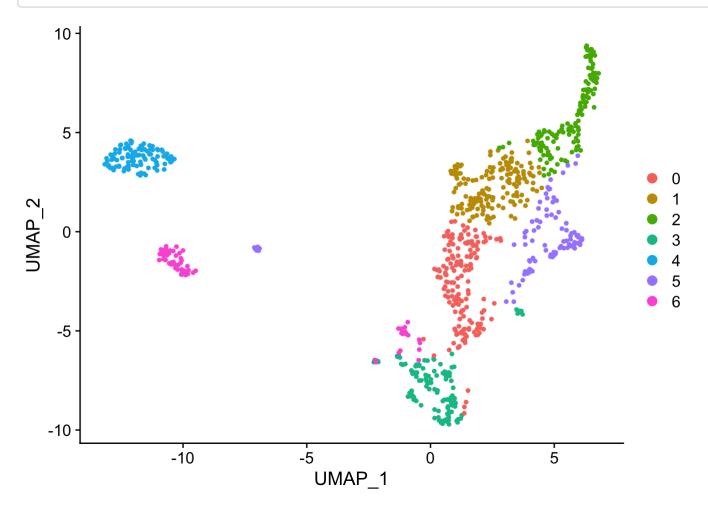
DimHeatmap (Neu_1k, dims = 1:15, cells = 50, balanced = TRUE)



Run UMAP on dataset

```
Neu_1k <- FindNeighbors(Neu_1k, dims = 1:15)</pre>
## Computing nearest neighbor graph
## Computing SNN
Neu_1k <- FindClusters (Neu_1k, resolution = 0.5)</pre>
## Modularity Optimizer version 1.3.0 by Ludo Waltman and Nees Jan van Eck
##
## Number of nodes: 862
## Number of edges: 23522
##
## Running Louvain algorithm...
## Maximum modularity in 10 random starts: 0.8617
## Number of communities: 7
## Elapsed time: 0 seconds
Neu_1k <- RunUMAP(Neu_1k, dims = 1:15)</pre>
## Warning: The default method for RunUMAP has changed from calling Python UMAP via reti
culate to the R-native UWOT using the cosine metric
## To use Python UMAP via reticulate, set umap.method to 'umap-learn' and metric to 'cor
relation'
## This message will be shown once per session
## 02:01:32 UMAP embedding parameters a = 0.9922 b = 1.112
## 02:01:32 Read 862 rows and found 15 numeric columns
## 02:01:32 Using Annoy for neighbor search, n neighbors = 30
## 02:01:32 Building Annoy index with metric = cosine, n_trees = 50
## 0%
       10
            20
                 30
                      40
                           50
                               60
                                     70
                                          80
                                               90
                                                    100%
## [----|----|----|
```

```
DimPlot (Neu_1k, reduction = "umap")
```



Find all markers of each cluster to identify them

```
cluster0.markers <- FindMarkers (Neu_1k, ident.1 = 0, min.pct = 0.5)
cluster1.markers <- FindMarkers (Neu_1k, ident.1 = 1, min.pct = 0.5)
cluster2.markers <- FindMarkers (Neu_1k, ident.1 = 2, min.pct = 0.5)
cluster3.markers <- FindMarkers (Neu_1k, ident.1 = 3, min.pct = 0.5)
cluster4.markers <- FindMarkers (Neu_1k, ident.1 = 4, min.pct = 0.5)
cluster5.markers <- FindMarkers (Neu_1k, ident.1 = 5, min.pct = 0.5)
cluster6.markers <- FindMarkers (Neu_1k, ident.1 = 6, min.pct = 0.5)</pre>
```

To view the genes lists on your RStudio console

View(cluster0.markers) View(cluster1.markers) View(cluster2.markers) View(cluster3.markers) View(cluster4.markers) View(cluster5.markers) View(cluster6.markers)

To "export" the genes lists to your current directory

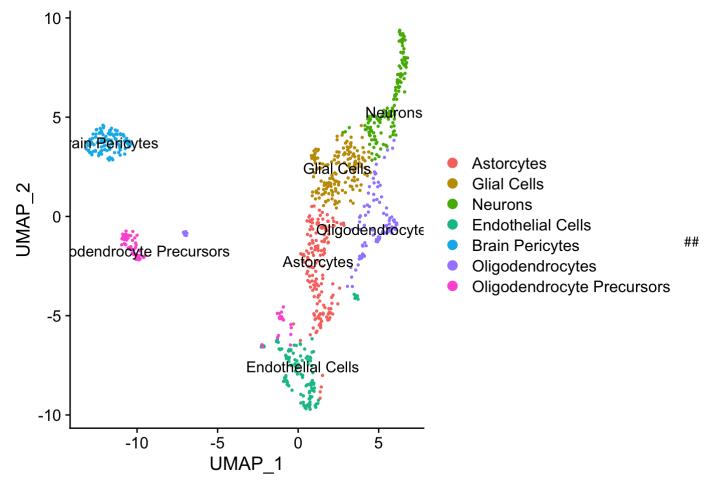
```
write.csv(cluster0.markers, "cluster0_markers_UMAP") write.csv(cluster1.markers, "cluster1_markers_UMAP") write.csv(cluster2.markers, "cluster2_markers_UMAP") write.csv(cluster3.markers, "cluster3_markers_UMAP") write.csv(cluster4.markers, "cluster4_markers_UMAP") write.csv(cluster5.markers, "cluster5_markers_UMAP") write.csv(cluster6.markers, "cluster6_markers_UMAP")
```

Find genes distinguishing one cluster from other specific clusters

```
green_cluster.markers <- FindMarkers (Neu_1k, ident.1 = 2, ident.2 = c(0, 1), min.pct =
0.25)</pre>
```

Label UMAP clusters with ids

```
new.cluster.ids <- c("Astorcytes", "Glial Cells", "Neurons", "Endothelial Cells", "Brain
Pericytes", "Oligodendrocytes", "Oligodendrocyte Precursors")
names(new.cluster.ids) <- levels (Neu_1k)
Neu_1k <- RenameIdents (Neu_1k, new.cluster.ids)
DimPlot (Neu_1k, reduction = "umap", label = TRUE, pt.size = 0.5)</pre>
```



Visualize feature expression

FeaturePlot(Neu_1k, features = c("Gria2", "Eomes", "Mef2c", "Fabp7", "Maf", "Snca", "Ada
rb2"))

