

# Results UMAP

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```
##### Required packages #####

library(dplyr)
library(Seurat)
library(cowplot)

##### Exp_1 #####

R048_1 <- Read10X(data.dir = "~/Dropbox/DataScience/Fiver/R048_1/")%>%
  CreateSeuratObject(min.cells = 3, min.features = 200,project = "R048 Exp.1",names.field

# FeatureScatter(object = R048_1, feature1 = "nCount_RNA", feature2 = "nFeature_RNA")

DMSO_1 <- Read10X(data.dir = "~/Dropbox/DataScience/Fiver/DMSO_1/")%>%
  CreateSeuratObject(min.cells = 3, min.features = 200,project = "DMSO Exp.1",names.field = 3)

# FeatureScatter(object = DMSO_1, feature1 = "nCount_RNA", feature2 = "nFeature_RNA")

mix_1 <- merge (x= DMSO_1,y=R048_1)%>%
  NormalizeData(verbose = FALSE)%>%
  FindVariableFeatures(selection.method = "vst", nfeatures = 2000)

#FeatureScatter(object = mix_1, feature1 = "nCount_RNA", feature2 = "nFeature_RNA",pt.size = 1,smo
# FeatureScatter(object = mix_1, feature1 = "nCount_RNA", feature2 = "nFeature_RNA",pt.size = 1,smo
rm(R048_1,DMSO_1)#remove objects to reduce RAM load

##### PCA exp_1 #####

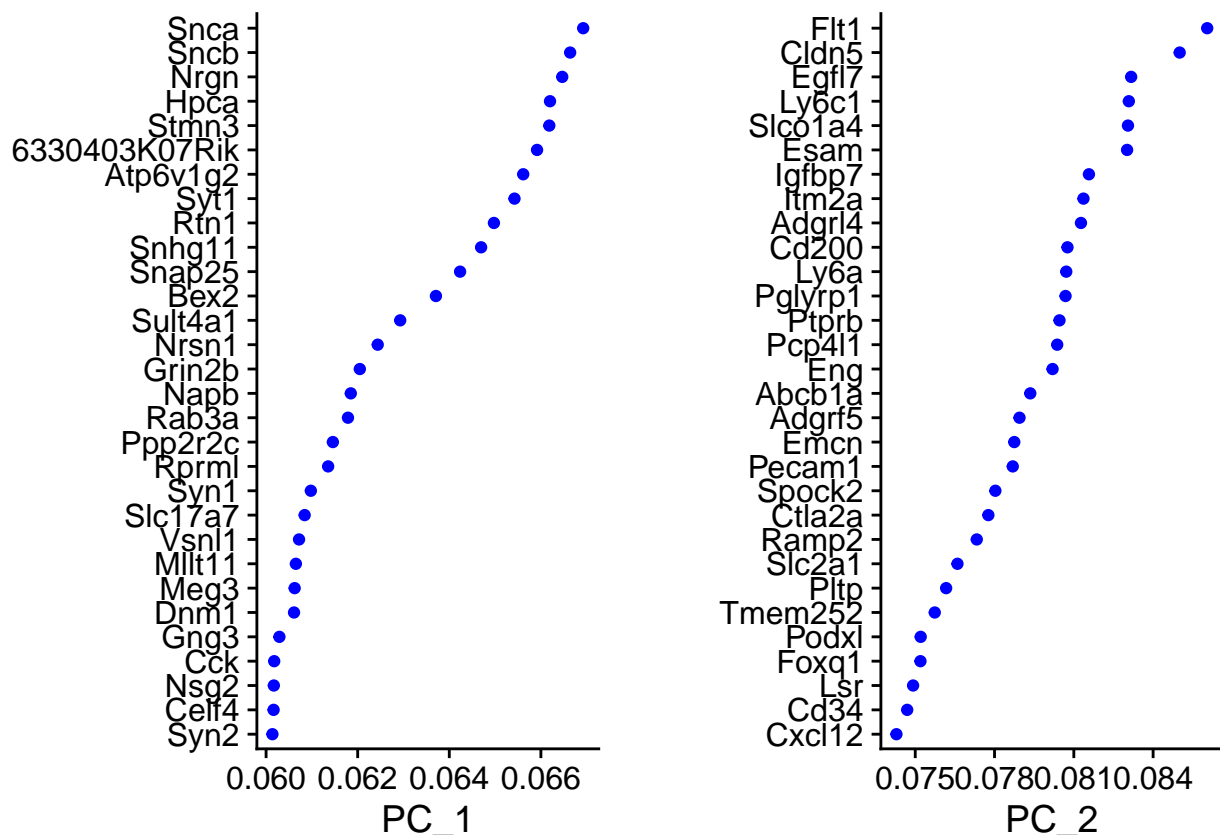
all.genes <- rownames(x = mix_1)

mix_1 <- ScaleData(object = mix_1, features = all.genes)
mix_1 <- RunPCA(object = mix_1, features = VariableFeatures(object = mix_1))
```

```
## PC_ 1
## Positive: Snca, Sncb, Nrgn, Hpca, Stmn3, 6330403K07Rik, Atp6v1g2, Syt1, Rtn1, Snhg11
## Snap25, Bex2, Sult4a1, Nrsn1, Grin2b, Napb, Rab3a, Ppp2r2c, Rprml, Syn1
## Slc17a7, Vsnl1, Mllt11, Meg3, Dnm1, Gng3, Cck, Nsg2, Celf4, Syn2
## Negative: Ftl1, Nfkb1a, Ifitm3, Sparc, Bcl2a1b, Ecscr, Fcgr3, Cd68, Ly86, Lmo2
## C1qb, C1qc, C1qa, Cd14, Lgmn, Hexb, Ifi30, Ctsd, Plin2, BC028528
## Selplg, Fcrls, Anxa2, Ms4a6d, Lilrb4a, Lyz2, Dab2, C5ar1, Pim1, Atf3
## PC_ 2
## Positive: Ftl1, Cldn5, Egfl7, Ly6c1, Slco1a4, Esam, Igfbp7, Itm2a, Adgrl4, Cd200
## Ly6a, Pglyrp1, Ptprb, Pcp4l1, Eng, Abcb1a, Adgrf5, Emcn, Pecam1, Spock2
## Ctla2a, Ramp2, Slc2a1, Pltp, Tmem252, Podxl, Foxq1, Lsr, Cd34, Cxcl12
## Negative: Mt1, Lgals1, Ctsd, Mt3, Gfap, Apoe, Aldoc, Fabp5, Nupr1, C1qb
## C1qc, C1qa, Cd68, Ly86, Tubb2b, Fcgr3, Bcl2a1b, Ndr2, Fabp7, Prdx6
## Clu, Slc1a2, S100b, Sdc4, Cd9, Lgmn, Fcrls, Ctsb, Dbi, Ifi30
```

```
## PC_ 3
## Positive: S100a16, Ptn, Dbi, Fabp7, Tubb2b, Gpr37l1, Atp1a2, Ptprz1, Mt3, Scd2
##           S100a1, Gpm6b, S100b, Ndr2, Aldoc, Atp1b2, Cspg5, Clu, Cnn3, Olig1
##           Tuba1a, Serpina3n, Cpe, Plpp3, Gfap, Slc1a2, Mmd2, Chst2, Htra1, Luzp2
## Negative: Ly86, Cd68, Fcgr3, C1qb, C1qc, Bcl2a1b, C1qa, Lgmn, Hexb, Ft11
##           Cd14, Fcrls, Ifi30, Cd83, Selp1g, Mef2c, Timp2, Ms4a6d, Cx3cr1, Lilrb4a
##           C5ar1, Tlr2, Lyz2, Gpr34, Ccl3, Fcgr1, Tmem119, Ccr12, Rgs1, Ccl4
## PC_ 4
## Positive: Gfap, Chchd10, Slc1a2, Clu, Aldoc, Mt3, Gstm1, Fxyd1, Nupr1, Cpe
##           Igfbp2, Fam107a, Acsbg1, Aldh1l1, Mlc1, Rbp1, Slc25a18, Apoe, Tst, Sparcl1
##           Aqp4, Prdx6, Gja1, Slc4a4, Timp1, Glul, Mrps6, Fgfr3, Ndr2, Lcn2
## Negative: Pdgra, Lhpl3, C1ql1, Sox10, Qpct, Cacng4, Tnr, Plp, Plppr5, Matn4
##           Pbk, 1700086L19Rik, Hmgn2, Gpr17, Nxph1, Lockd, Vcan, Sulf2, Olig2, Marcks11
##           2810417H13Rik, Bcas1, Cdca3, Lims2, Top2a, Ugdh, Dpysl3, H2afz, Lrrn1, Sema3d
## PC_ 5
## Positive: Top2a, 2810417H13Rik, Birc5, Pbk, Ccna2, Cdca3, Mki67, Cks1b, Tubb5, Prc1
##           Nusap1, Cdca8, Ube2c, Spc25, Cdk1, Racgap1, Smc2, Tpx2, Smc4, Cenpf
##           F3, Slc1a3, Tuba1b, Rrm2, Ccnb1, Spc24, Hmnr, Knstrn, Cenpe, Mt2
## Negative: Mog, Opalin, Nkx6-2, Ernm, Mag, Cldn11, Plp1, Klk6, Ptgs, Mobp
##           Gjb1, Fa2h, Tmem88b, Tspan2, Ppp1r14a, Tmem125, Aspa, Gjc2, Efnb3, Tubb4a
##           Ugt8a, Hapln2, Mbp, Mal, Apod, Tnni1, Lgi3, Grb14, Tmeff2, Stmn4
```

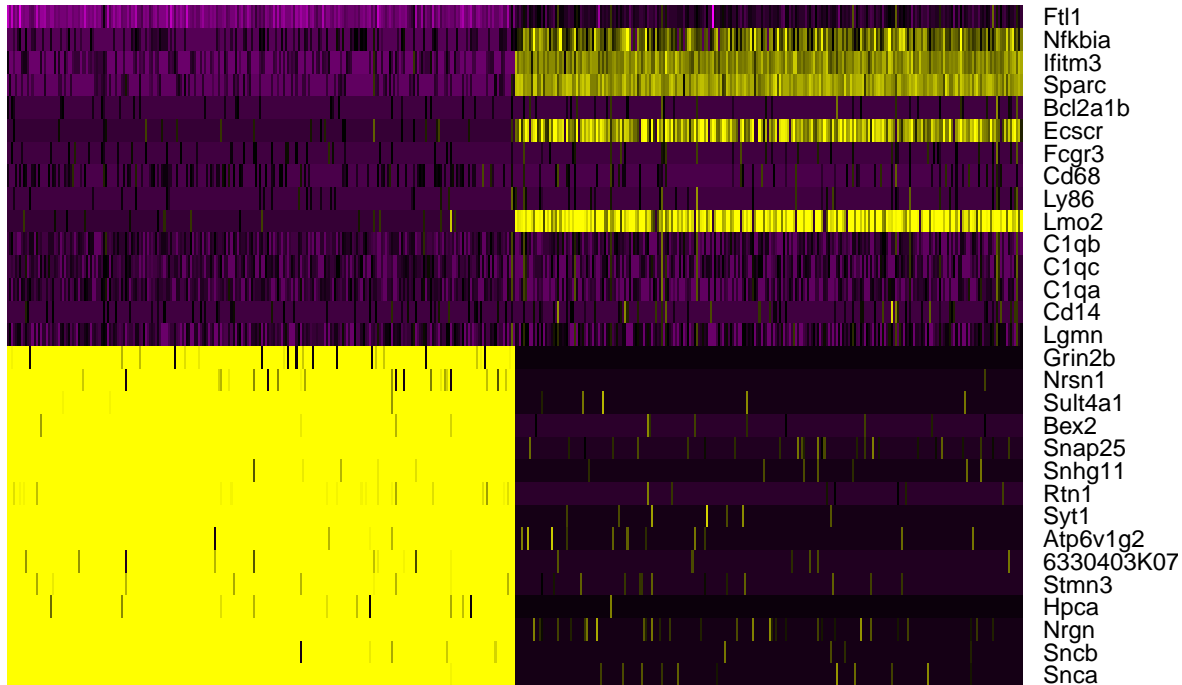
```
VizDimLoadings(object = mix_1, dims = 1:2, reduction = 'pca')
```



```
# DimPlot(object = mix_1, reduction = 'pca')
```

```
DimHeatmap(object = mix_1, dims = 1, cells = 500, balanced = TRUE)
```

## PC\_1



```
# DimHeatmap(object =mix_1, dims = 1:15, cells = 500, balanced = TRUE)
```

```
# ElbowPlot(object = mix_1)
```

```
##### Clusters exp_1 #####
```

```
mix_1 <- FindNeighbors(object = mix_1, dims = 1:10)
```

```
mix_1 <- FindClusters(object = mix_1, resolution = 0.5)
```

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

```
##
```

```
## Number of nodes: 22563
```

```
## Number of edges: 736253
```

```
##
```

```
## Running Louvain algorithm...
```

```
## Maximum modularity in 10 random starts: 0.9422
```

```
## Number of communities: 21
```

```
## Elapsed time: 3 seconds
```

```
# find all markers of cluster 1
```

```
# cluster1.markers <- FindMarkers(object = mix_1, ident.1 = 1, min.pct = 0.25)
```

```
# find all markers distinguishing cluster 5 from clusters 0 and 3
```

```
# cluster5.markers <- FindMarkers(object = mix_1, ident.1 = 5, ident.2 = c(0, 3), min.pct = 0.25)
```

```
# find markers for every cluster compared to all remaining cells, report only the positive ones
```

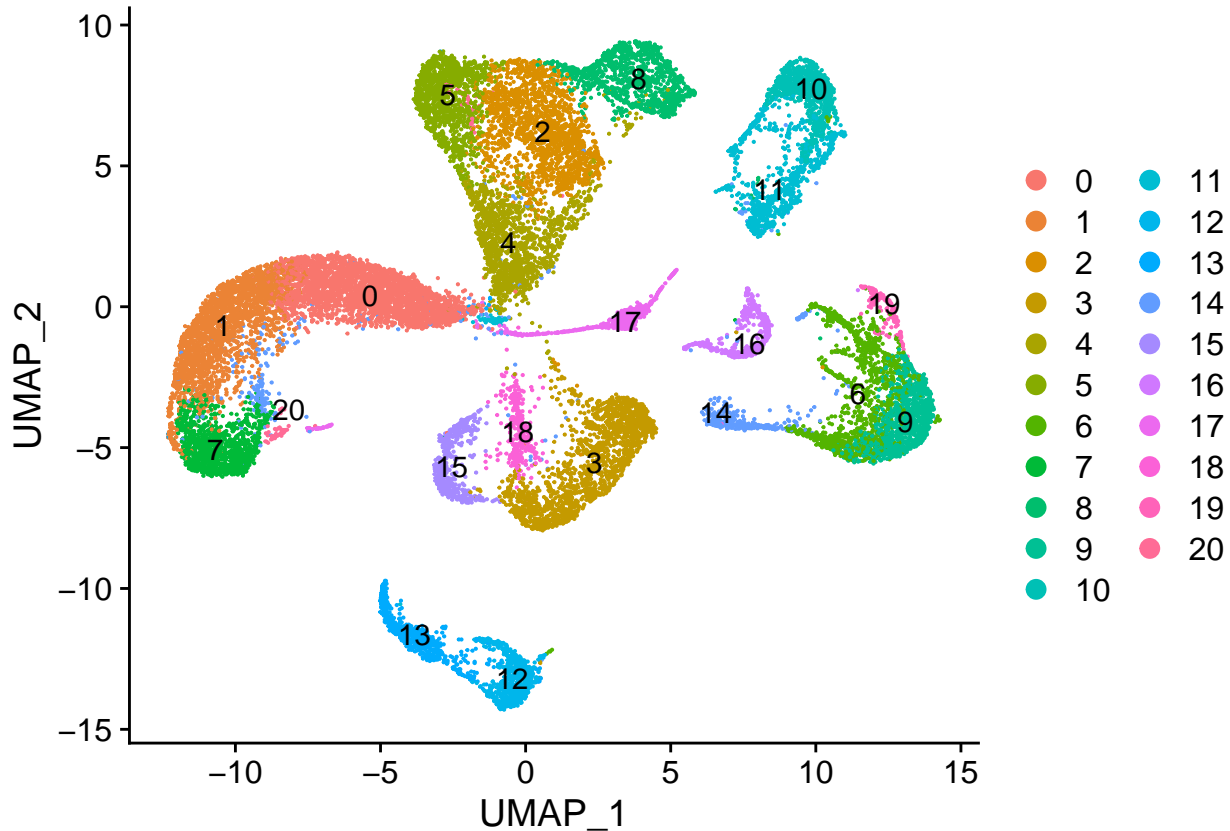
```
# pbmc.markers <- FindAllMarkers(object = mix_1, only.pos = TRUE, min.pct = 0.25, logfc.threshold = 1)
```

```
# pbmc.markers %>% group_by(cluster) %>% top_n(n = 2, wt = avg_logFC)
```

```
##### UMAP exp_1 #####
#Need to install pip on Mac and run in console: sudo apt-get install pip
#Install the package with: pip install umap-learn
```

```
mix_1 <- RunUMAP(object = mix_1, dims = 1:10)

DimPlot(object = mix_1, reduction = 'umap',label = T)
```



```
#####
```

```
rm(mix_1,all.genes)
##### Exp_2 #####

R048_2 <- Read10X(data.dir = "~/Dropbox/DataScience/Fiver/R048_2/")%>%
  CreateSeuratObject(min.cells = 30, min.features = 2000,project = "R048 Exp.2")

# FeatureScatter(object = R048_2, feature1 = "nCount_RNA", feature2 = "nFeature_RNA")

DMSO_2 <- Read10X(data.dir = "~/Dropbox/DataScience/Fiver/DMSO_2/")%>%
  CreateSeuratObject(min.cells = 30, min.features = 2000,project = "DMSO Exp.2")

# FeatureScatter(object = DMSO_2, feature1 = "nCount_RNA", feature2 = "nFeature_RNA")

mix_2 <- merge (x= DMSO_2,y=R048_2)%>%
  NormalizeData(verbose = FALSE)%>%
  FindVariableFeatures(selection.method = "vst", nfeatures = 2000)

# FeatureScatter(object = mix_2, feature1 = "nCount_RNA", feature2 = "nFeature_RNA",pt.size = 1,sm
```

```

# FeatureScatter(object = mix_2, feature1 = "nCount_RNA", feature2 = "nFeature_RNA", pt.size = 1, sm
rm(RO48_2,DMSO_2)#remove objects to reduce RAM load

##### PCA exp_2 #####

all.genes <- rownames(x = mix_2)

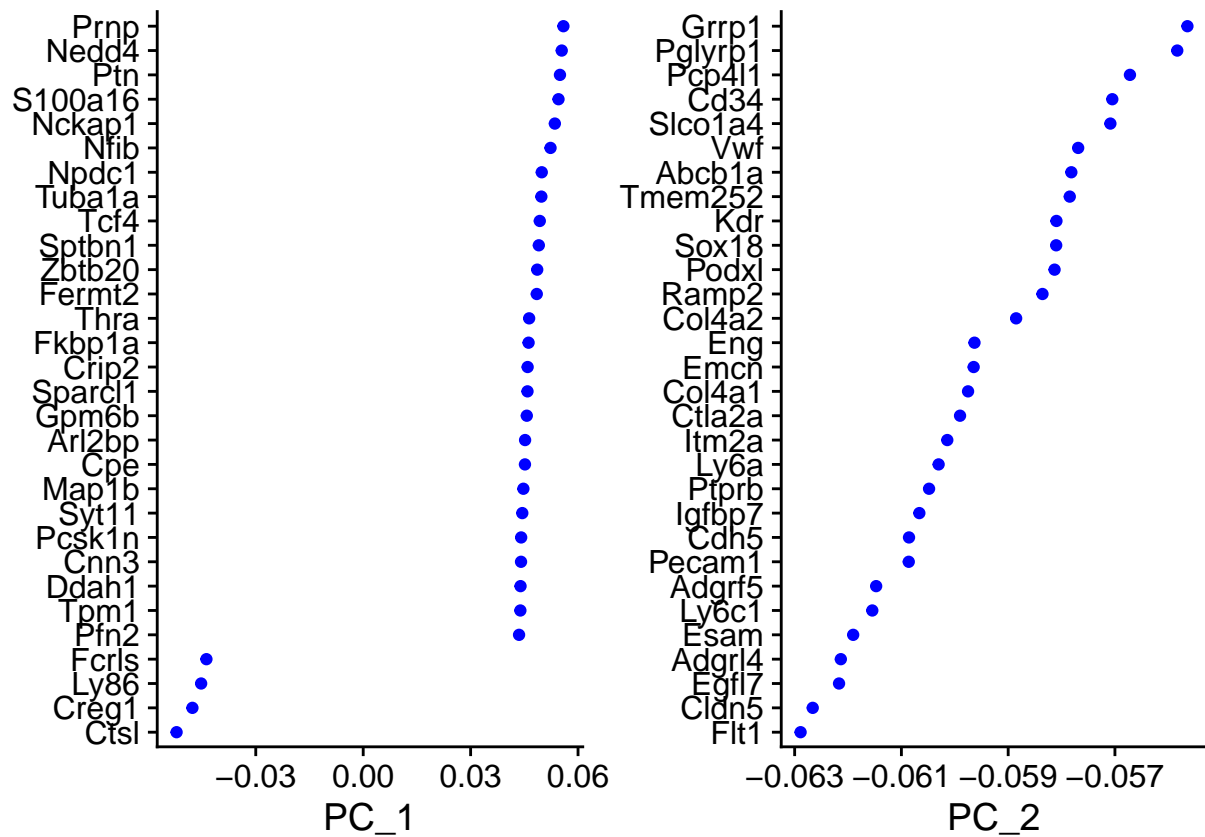
mix_2 <- ScaleData(object = mix_2, features = all.genes)
rm(all.genes)

mix_2 <- RunPCA(object = mix_2, features = VariableFeatures(object = mix_2))

## PC_ 1
## Positive: Prnp, Nedd4, Ptn, Slc17a7, Nkap1, Nfib, Npdc1, Tuba1a, Tcf4, Sptbn1
##           Zbtb20, Fermt2, Thra, Fkbp1a, Crip2, Sparcl1, Gpm6b, Arl2bp, Cpe, Map1b
##           Syt11, Pcsk1n, Cnn3, Ddah1, Tpm1, Pfn2, Tsc22d1, Scd2, Selenom, Uchl1
## Negative: Ctsl, Creg1, Ly86, Fcrls, Lyz2, Plin2, Psap, Rgs1, Olflm3, Cd14
##           Zfp36, Cx3crl, Atf3, Lgals3, Spp1, Wfdc17, Msr1, Id2, Apoe, Ccrl2
##           Tgfb1, Tmem119, Ccl3, Rgs2, Ifi2712a, Ccl4, Cd83, Npl, Ms4a7, Gpr34
## PC_ 2
## Positive: Syt11, Ckb, Gpm6b, Pcsk1n, Anks1b, Mapt, Kif1a, Ank2, Gnao1, Ank3
##           Olig1, Aplp1, Fez1, Pcdh9, Ncam1, Shisa4, Clip3, Elavl3, Fam171b, Tmod2
##           Sv2a, Serpina3n, Trim2, Pfn2, Slc24a2, Cnp, Tubb4a, Ncald, Phyhipl, Stmn4
## Negative: Flt1, Cldn5, Egfl7, Adgrl4, Esam, Ly6c1, Adgrf5, Pecam1, Cdh5, Igfbp7
##           Ptprb, Ly6a, Itm2a, Ctla2a, Col4a1, Emcn, Eng, Col4a2, Ramp2, Podxl
##           Sox18, Kdr, Tmem252, Abcb1a, Vwf, Slc1a4, Cd34, Pcp4l1, Pglyrp1, Grp1
## PC_ 3
## Positive: Syt1, Slc17a7, Snap25, Snhg11, Dnm1, Myt11, Grin2b, Camk2b, Syn2, Adcy1
##           Ndrp4, Scn2a, Grin1, Slc6a17, Sv2b, Celf4, Atp6v1g2, Plppr4, Rasgrp1, Thy1
##           Hpcal4, Nrnx3, Slc4a10, Nptx1, Sult4a1, Syp, Pde1a, Gabra1, Pcsk2, Vsnl1
## Negative: Mog, Mag, Cldn11, Nkx6-2, Ernm, Mobp, Ppp1r14a, Plp1, Tmem88b, Tspan2
##           Car2, Opalin, Mal, Ugt8a, Sept4, Efnb3, Gpr37, Cnp, Mbp, Apod
##           Trf, Enpp2, Aspa, Gjb1, Gjc2, Fa2h, Lgi3, Grb14, Tubb4a, Klk6
## PC_ 4
## Positive: Slc24a2, Syt1, Slc17a7, Spock1, Dnm1, Grin2b, Snap25, Tmem151a, Sv2b, Srcin1
##           Snhg11, Hpcal4, Adcy1, Dnajc6, Grin1, Syp, Ppp2r2c, Sv2a, Eno2, Nrgn
##           Rbfox3, Myt11, Camk2a, Gabra1, Napb, Camk2b, Snca, Sult4a1, Pacsin1, Nptx1
## Negative: Ptprz1, Fabp7, Gpr3711, Tubb2b, Atp1b2, Cspg5, Atp1a2, F3, Gm3764, Slc1a3
##           Tril, Luzp2, Mt3, Ndrp2, Timp4, Ednrb, Aqp4, Mlc1, Tmem47, Scrg1
##           Slc4a4, Acsbg1, Fgfr3, Mmd2, Mdk, Igfbp2, Sox9, Clu, Ntrk2, Mtss11
## PC_ 5
## Positive: Pla2g7, Prdx6, Mgst1, Chchd10, Gstm1, Fabp5, Slc4a4, Mlc1, Apoe, Psap
##           Aqp4, Clu, Gpnmb, Thbs1, Slc6a11, Aldoc, Fxyd1, Ecml, Vegfa, Fgfr3
##           Clec4d, Lgals3, Lyz2, Gjb6, Aldh1l1, Cldn10, Acsl6, Fam107a, Acsbg1, Gfap
## Negative: Pclaf, Birc5, Top2a, Mki67, Tubb5, Stmn1, Cdk1, Pbk, Ube2c, Ccna2
##           Cdca3, H2afz, Cdca8, Prc1, Tuba1b, Smc2, Tpx2, Cenpf, Racgap1, Cks1b
##           Cks2, Selenoh, Ccnb1, Tmpo, Spc24, Nusap1, Rrm2, Hmgb2, H2afx, Tk1

VizDimLoadings(object = mix_2, dims = 1:2, reduction = 'pca')

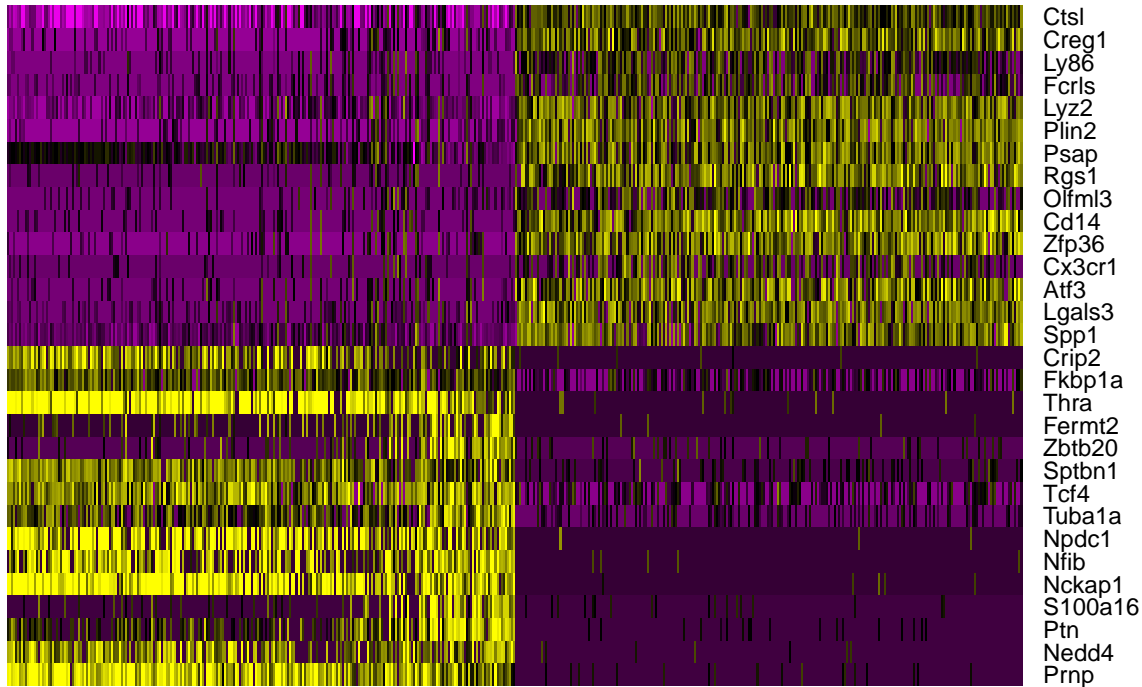
```



```
# DimPlot(object = mix_2, reduction = 'pca')
```

```
DimHeatmap(object = mix_2, dims = 1, cells = 500, balanced = TRUE)
```

## PC\_1



```
# DimHeatmap(object =mix_2, dims = 1:15, cells = 500, balanced = TRUE)
```

```
# ElbowPlot(object = mix_2)
```

```
##### Clusters exp_2 #####
```

```
mix_2 <- FindNeighbors(object = mix_2, dims = 1:10)
```

```
mix_2 <- FindClusters(object = mix_2, resolution = 0.5)
```

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

```
##
```

```
## Number of nodes: 19093
```

```
## Number of edges: 603438
```

```
##
```

```
## Running Louvain algorithm...
```

```
## Maximum modularity in 10 random starts: 0.9207
```

```
## Number of communities: 17
```

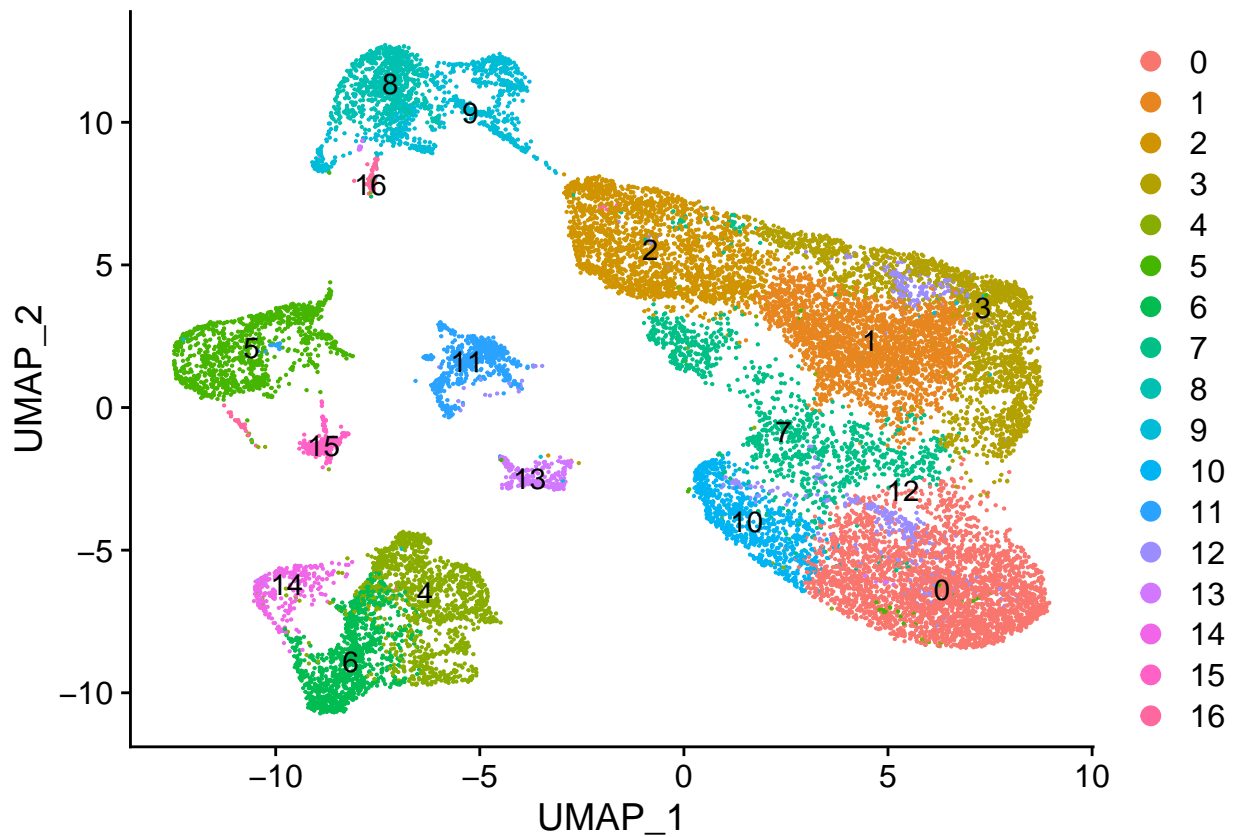
```
## Elapsed time: 2 seconds
```

```
##### UMAP exp_2 #####
```

```
#pip install umap-learn
```

```
mix_2 <- RunUMAP(object = mix_2, dims = 1:10)
```

```
DimPlot(object = mix_2, reduction = 'umap',label = T)
```



```
##### cluster biomarkers exp_2 #####

# find all markers of cluster 1

#cluster1.markers <- FindMarkers(object = mix_2, ident.1 = 1, min.pct = 0.25)

# find all markers distinguishing cluster 5 from clusters 0 and 3

#cluster5.markers <- FindMarkers(object = mix_2, ident.1 = 5, ident.2 = c(0, 3), min.pct = 0.25)

# find markers for every cluster compared to all remaining cells, report only the positive ones

# pbmc.markers <- FindAllMarkers(object = mix_2, only.pos = TRUE, min.pct = 0.25, logfc.threshold = 1)
# pbmc.markers %>% group_by(cluster) %>% top_n(n = 2, wt = avg_logFC)
```