## SCMarker.NK.GBM

library(cowplot)

The GBM data set is downloaded from GEO (GSE84465), including 3587 cells from four primary GBM patients. Darmanis S, Sloan SA, Croote D, Mignardi M et al. Single-Cell RNA-Seq Analysis of Infiltrating Neoplastic Cells at the Migrating Front of Human Glioblastoma. Cell Rep 2017 Oct 31;21(5):1399-1410. PMID: 29091775

```
GBM=read.csv('GSE84465 GBM All data.csv',sep="")
 dim(GBM)
 ## [1] 23465 3589
Run Seurat to identify cluster enriched NK cell
 library(dplyr)
 ## Warning: package 'dplyr' was built under R version 3.5.2
 ##
 ## 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)
```

```
## Warning: package 'cowplot' was built under R version 3.5.2
##
## *******************
## Note: As of version 1.0.0, cowplot does not change the
##
    default ggplot2 theme anymore. To recover the previous
##
    behavior, execute:
##
    theme set(theme cowplot())
## ********************
library(ggplot2)
## Warning: package 'ggplot2' was built under R version 3.5.2
GBMobj <- CreateSeuratObject(counts = GBM, project = "GBM")</pre>
## Warning: Feature names cannot have underscores ('_'), replacing w
ith dashes
## ('-')
GBMobj <- NormalizeData(object = GBMobj, verbose = FALSE)</pre>
GBMobj <- FindVariableFeatures(object = GBMobj, selection.method = "</pre>
vst", nfeatures = 2000)
GBMobj <- ScaleData(GBMobj)</pre>
## Centering and scaling data matrix
GBMobj <- RunPCA(GBMobj)</pre>
```

```
## PC 1
## Positive: GPM6B, PTPRZ1, SCD5, GPRC5B, TUBA1A, TSPAN7, GPM6A, CL
U, BCAN, NCAM1
##
       ALDOC, SLC22A17, CPE, DNER, TSC22D1, S100B, SERPINE2, SPARCL1
, NLGN3, FXYD6
##
       Clorf61, BAALC, CNP, PLEKHB1, OLIG1, GPR37L1, PTN, SPOCK1, AP
OD, SLC1A2
## Negative: CD74, HLA-DRA, HLA-DRB1, C1QB, RGS1, HLA-DPA1, CD83, O
LR1, HLA-DRB5, CCL3
##
      GPR183, A2M, CSF1R, HLA-DPB1, IFI30, FTL, C3, CTSB, VSIG4, FP
R1
##
      CCL4, SPP1, CD14, HLA-DQB1, PLEK, PLAUR, HLA-DMB, ACSL1, APOE
, RHBDF2
## PC 2
## Positive: TF, DBNDD2, MOG, TUBB4A, ENPP2, ERMN, PPP1R14A, PLLP,
KLK6, SPOCK3
##
      CNDP1, CNTN2, EDIL3, PLP1, MAL, C11orf9, APOD, OPALIN, CARNS1
, MBP
##
      MAG, TMEM144, MOBP, UGT8, CNP, HAPLN2, FA2H, SPOCK1, SH3GL3,
SEPT4
## Negative: IGFBP7, IGFBP2, TNC, CYR61, CSRP2, C1R, WWTR1, CLU, SE
RPINA3, AQP4
##
       CHI3L1, GFAP, ID3, MT2A, ID4, C8orf4, ACTN1, GADD45A, MLC1, C
1S
##
      HSPB1, CA12, GAP43, EGFR, COL1A2, EMP1, CAV1, SPARC, CXCL14,
GEM
## PC 3
## Positive: TNR, MEGF11, MEG3, ATCAY, PDGFRA, SMOC1, NKAIN4, GPR17
, CA10, CSPG4
##
      SULF2, PDZD2, EPN2, SEZ6L, OLIG1, EHD3, PCDH15, LHFPL3, DLL3,
NLGN3
##
      GFRA1, AFAP1L2, NXPH1, BCAN, VCAN, COL9A1, BAMBI, FLJ43390, H
IP1R, FXYD6
## Negative: DBNDD2, ERMN, KLK6, CNDP1, TF, PLP1, CRYAB, CNTN2, TME
M144, MOG
##
      MOBP, ENPP2, MAG, OPALIN, CARNS1, PPP1R14A, SEPT4, MAL, GPR37
, CAPN3
##
      EFHD1, ANLN, C11orf9, MBP, LPAR1, SEPP1, ABCA8, FOLH1, SH3GL3
, CLDND1
## PC 4
## Positive: TIMP1, ANXA2, CAV1, GAP43, C8orf4, MAP1B, IGFBP2, TAGL
N, FN1, AKAP12
```

```
##
       TPM2, GADD45A, CALD1, WWTR1, SPOCD1, CA9, MGP, THY1, ACTN1, C
OL1A2
##
       TM4SF1, IGFBP3, GRB10, RND3, CA12, CYR61, SCG2, MLLT11, IGFBP
5, EMP1
## Negative: AGXT2L1, GJB6, GPR98, WIF1, PPP1R1B, NTSR2, SLC39A12,
GABRA2, ACSBG1, SLC14A1
       CPNE6, PRODH, CLDN10, HHATL, GJA1, SPON1, ATP1A2, SLC1A3, SLC
1A4, SLC1A2
##
       AQP4, GABRB1, RYR3, FAM107A, AGT, SLC6A11, NDRG2, TPD52L1, GP
C5, HSD17B6
## PC 5
## Positive: IFITM1, DCN, COL3A1, NID1, GGT5, COL1A1, FRZB, MYO1B,
CD248, COL4A1
##
       PDGFRB, APOLD1, COL1A2, CFH, ISLR, COL4A2, COL6A2, IGFBP4, UA
CA, IGFBP7
       PCOLCE, FBLN2, COLEC12, LUM, FBLN1, OLFML3, MGLL, NDUFA4L2, C
##
YP1B1, ITIH5
## Negative:
            KCNE1L, DBI, Clorf61, HES6, IL13RA2, HIST1H4C, CCND2,
CREB5, EGFR, NNAT
       TUBA1A, CD9, STMN1, NREP, CENPF, MEST, CDK4, MASP1, PMP2, UBE
##
2T
##
       CD24, AQP4, NMB, FABP7, MLLT11, BIRC5, PBK, RARRES2, MKI67, U
BE2C
```

```
GBMobj <- FindNeighbors(GBMobj, dims = 1:25)</pre>
```

## Computing nearest neighbor graph

## Computing SNN

GBMobj <- FindClusters(GBMobj, resolution = 1.7)</pre>

```
## Modularity Optimizer version 1.3.0 by Ludo Waltman and Nees Jan v
an Eck
##
## Number of nodes: 3589
## Number of edges: 114422
##
## Running Louvain algorithm...
## Maximum modularity in 10 random starts: 0.8378
## Number of communities: 28
## Elapsed time: 0 seconds
GBMobj <- RunUMAP(GBMobj, dims = 1:25)</pre>
## Warning: The default method for RunUMAP has changed from calling
Python UMAP via reticulate to the R-native UWOT using the cosine met
ric
## To use Python UMAP via reticulate, set umap.method to 'umap-learn
' and metric to 'correlation'
## This message will be shown once per session
## 15:40:07 UMAP embedding parameters a = 0.9922 b = 1.112
## 15:40:07 Read 3589 rows and found 25 numeric columns
## 15:40:07 Using Annoy for neighbor search, n neighbors = 30
## 15:40:07 Building Annoy index with metric = cosine, n trees = 50
## 0%
       10
            20
                 30
                      40
                           50
                                60
                                     70
                                         80
                                              90
                                                   100%
## [----|----|----|
```

```
## *****************************

## 15:40:07 Writing NN index file to temp file /var/folders/gm/1x2dv
vj50wvd8ncvyqm5z6ssyh7m9q/T//RtmpXuw6Kd/filee28558c0d09d

## 15:40:07 Searching Annoy index using 1 thread, search_k = 3000

## 15:40:08 Annoy recall = 100%

## 15:40:08 Commencing smooth kNN distance calibration using 1 thread
d

## 15:40:08 Initializing from normalized Laplacian + noise

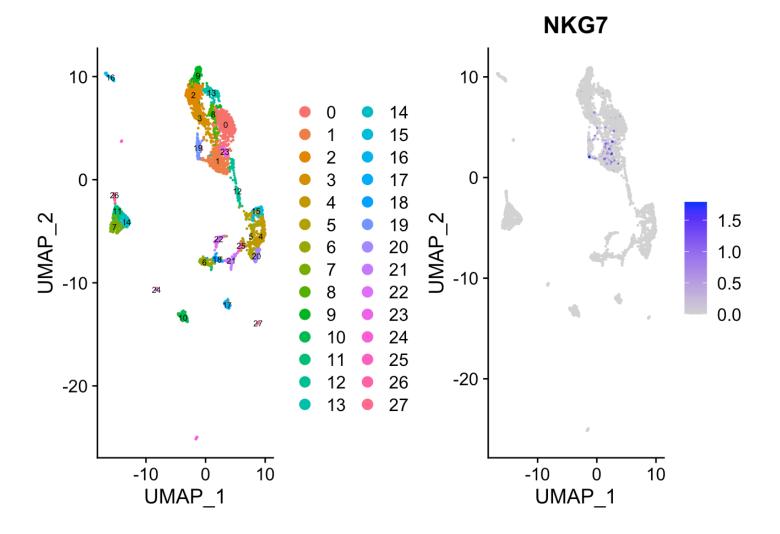
## 15:40:09 Commencing optimization for 500 epochs, with 140114 positive edges

## 15:40:13 Optimization finished
```

```
p1 <- DimPlot(GBMobj, label = TRUE, pt.size = 0.2, label.size=2)</pre>
```

```
## Warning: Using `as.character()` on a quosure is deprecated as of
rlang 0.3.0.
## Please use `as_label()` or `as_name()` instead.
## This warning is displayed once per session.
```

```
p2 <- FeaturePlot(GBMobj, features = c("NKG7"),pt.size=0.2)
p1+p2</pre>
```

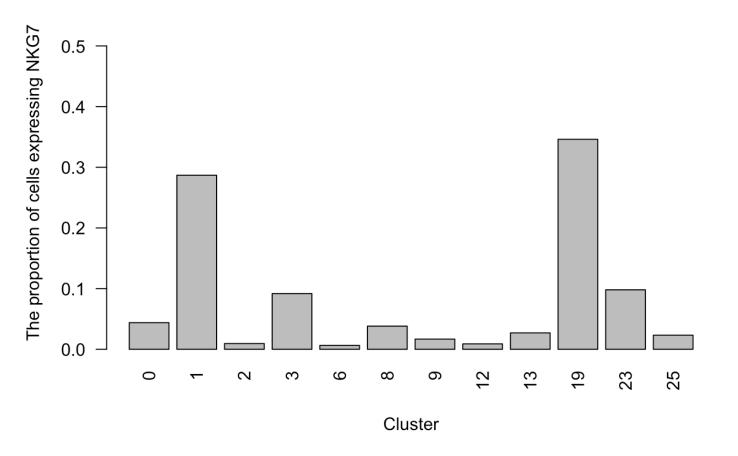


The proportion of cells expressing NK cell marker NKG7 in each cluster

```
subGBM=subset(GBMobj,NKG7>0)
index=match(names(table(Idents(subGBM))),names(table(Idents(GBMobj))
))
seurat=table(Idents(subGBM))/table(Idents(GBMobj))[index]
```

```
barplot(seurat,ylim = c(0,0.55),las=2,main='Seurat',xlab="Cluster",y lab="The proportion of cells expressing NKG7")
```

## Seurat



## Integration of Seurat and SCMarker

```
library(SCMarker)
GBM1=log(GBM+1)
scMarker.res=ModalFilter(data=GBM1,geneK=10,cellK=10,width=2)# defau
lt width = 1 for UMI data, width =2 for TPM data.
scMarker.res=GeneFilter(obj=scMarker.res)
scMarker.res=getMarker(obj=scMarker.res,k=300,n=30)
```

## Extract markers identified by SCMarker

```
scmarker=scMarker.res$marker
length(scmarker)

## [1] 840
```

```
head(scmarker)
```

```
## [1] "ABL2" "ACSL1" "ADORA3" "AIF1" "AKR1B1" "ALOX5"
all.genes <- rownames(GBMobj)</pre>
GBMscmarker <- ScaleData(GBMobj,features = all.genes)</pre>
## Centering and scaling data matrix
###Run PCA based on union of markers from SCMarker and Seurat
GBMscmarker <- RunPCA(GBMscmarker, features = union(scmarker, Variabl
eFeatures(GBMobj)))
## PC 1
## Positive: CD74, LAPTM5, SRGN, HLA-DRA, HLA-DRB1, C1QB, TYROBP, C
1QA, FCER1G, RNASET2
##
       SLC11A1, RGS1, HLA-DPA1, AIF1, CD83, OLR1, HLA-DRB5, CCL3, GP
R183, HLA-DPB1
##
      CSF1R, A2M, IFI30, GPX1, ARHGDIB, FTL, HLA-DMA, C3, VSIG4, HC
LS1
## Negative: PTPRZ1, GPM6B, TUBA1A, CLU, CNN3, GPM6A, NGFRAP1, SCD5
, MARCKSL1, TSPAN7
##
      Clorf61, BCAN, GPRC5B, SYT11, PTN, NRCAM, CPE, FEZ1, ITGB8, L
DHB
##
      ALDOC, BAALC, FXYD6, MAP1B, SRI, PEBP1, TRIM9, PON2, SLC22A17
, TSPAN3
## PC 2
## Positive: VIM, ANXA2, IGFBP2, LDHA, CHCHD2, CD44, IGFBP7, TPI1,
CSRP2, TAGLN2
      LGALS3, ACTN1, WWTR1, TNC, ANXA1, S100A10, CALU, MT2A, GAPDH,
##
C1R
##
      CYR61, CA12, CAV1, SOD2, TUBA1C, TIMP1, TUBA1B, CHI3L1, SERPI
NA3, CCT6A
## Negative: APOD, PLLP, SOX10, SPOCK1, TNR, CDR1, SNX22, MEGF11, S
MOC1, EDIL3
##
     GPR17, ATCAY, OMG, PDGFRA, MEG3, NKAIN4, OLIG1, TUBB4A, EPN2,
HIP1R
##
     CA10, BCAS1, PCDH15, BAMBI, EHD3, FAIM2, CNP, PDZD2, CSPG4, S
OX8
## PC 3
```

## Positive: HTRA1, COTL1, CTSD, BIN1, CXCL16, GRN, CTSB, FPR1, COR

```
O1A, GPX1
##
       TSPO, SERPINA1, YWHAH, CAPG, SPP1, UCP2, C1QB, TREM2, CREG1,
FAM49B
      LAPTM5, CD14, CNP, CNDP2, CD81, RHOA, ITGB2, ARF1, C20orf3, S
##
GK1
## Negative: ERCC-00131, GJA1, AQP4, ERCC-00071, ID4, ERCC-00044, R
YR3, ERCC-00079, HSPB1, GFAP
      ERCC-00112, ERCC-00116, GPR98, ERCC-00025, ERCC-00144, Clorf6
1, ID3, S1PR1, AGT, AGXT2L1
      CYR61, EMP1, ERCC-00062, C8orf4, DTNA, CHI3L1, ABCA13, RARRES
##
2, SULF1, CLU
## PC 4
## Positive: DBNDD2, ERMN, TF, KLK6, CNDP1, MOG, ENPP2, PLP1, CNTN2
, PPP1R14A
##
       TMEM144, MOBP, OPALIN, CARNS1, MAG, MAL, SEPT4, CAPN3, C11orf
9, MBP
##
       SH3GL3, ANLN, EFHD1, QDPR, FOLH1, APLP1, RAPGEF5, CRYAB, TPPP
, ST18
## Negative: MEG3, TNR, ATCAY, MEGF11, PDGFRA, CSPG4, BCAN, PDZD2,
LHFPL3, SULF2
##
     NKAIN4, SMOC1, CA10, SEZ6L, FXYD6, PTPRZ1, KCNQ2, CST3, GPR17
, EHD3
##
      VCAN, PCDH15, C1QL1, TSPAN7, XYLT1, EPN2, CACNG4, SEMA5A, OLI
G1, GPR56
## PC 5
## Positive: AGXT2L1, GJB6, GPR98, SLC1A3, WIF1, PPP1R1B, GJA1, NTS
R2, NDRG2, SLC39A12
       FAM107A, ACSBG1, GABRA2, MLC1, SLC14A1, PRODH, CLDN10, CPNE6,
##
CPE, RYR3
##
      HHATL, DKK3, SPON1, GABRB1, TTYH1, TPD52L1, SPARCL1, AQP4, SL
C6A11, SLC1A2
## Negative: STMN1, HIST1H4C, CREB5, NPM1, TM4SF1, ATP50, S100A10,
RAN, HINT1, GNB2L1
##
      MLLT11, LDHA, LIMA1, IL13RA2, MDK, TGFBI, PSMA7, ANXA2, PDLIM
1, MKI67
##
      VCAN, CENPF, TIMP1, FXYD6, CACYBP, SH3BGRL3, KCNE1L, EEF1G, B
TF3, GST01
```

```
## Computing nearest neighbor graph
## Computing SNN
GBMscmarker <- FindClusters(GBMscmarker, resolution = 1.7)
## Modularity Optimizer version 1.3.0 by Ludo Waltman and Nees Jan v
an Eck
##
## Number of nodes: 3589
## Number of edges: 112377
##
## Running Louvain algorithm...
## Maximum modularity in 10 random starts: 0.8473
## Number of communities: 28
## Elapsed time: 0 seconds
GBMscmarker <- RunUMAP(GBMscmarker, dims = 1:25)
## 15:48:41 UMAP embedding parameters a = 0.9922 b = 1.112
## 15:48:41 Read 3589 rows and found 25 numeric columns
## 15:48:41 Using Annoy for neighbor search, n neighbors = 30
## 15:48:41 Building Annoy index with metric = cosine, n trees = 50
## 0% 10 20
                 30 40 50 60 70 80 90 100%
## [----|----|----|
```

```
## ****************************

## 15:48:41 Writing NN index file to temp file /var/folders/gm/1x2dv
vj50wvd8ncvyqm5z6ssyh7m9q/T//RtmpXuw6Kd/filee2851ba4fed3

## 15:48:41 Searching Annoy index using 1 thread, search_k = 3000

## 15:48:42 Annoy recall = 100%

## 15:48:42 Commencing smooth kNN distance calibration using 1 thread

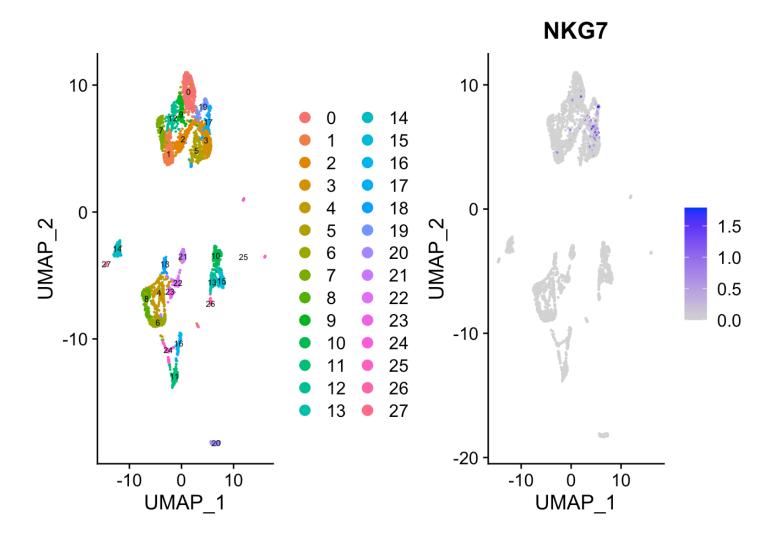
d

## 15:48:43 Initializing from normalized Laplacian + noise

## 15:48:43 Commencing optimization for 500 epochs, with 138724 posi
tive edges

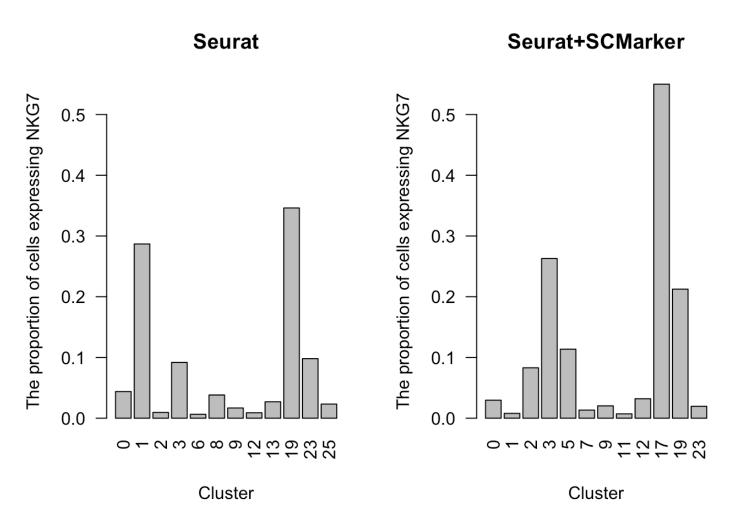
## 15:48:47 Optimization finished
```

```
p1 <- DimPlot(GBMscmarker, label = TRUE, pt.size = 0.2, label.size=2)
p2 <- FeaturePlot(GBMscmarker, features = c("NKG7"), pt.size=0.2)
p1+p2</pre>
```



```
subGBMmarker=subset(GBMscmarker,NKG7>0)
index=match(names(table(Idents(subGBMmarker))),names(table(Idents(GB
Mscmarker))))
scmarkercluster=table(Idents(subGBMmarker))/table(Idents(GBMscmarker
))[index]
```

```
par(mfrow=c(1,2))
barplot(seurat,ylim = c(0,0.55),las=2,main='Seurat',xlab="Cluster",y
lab="The proportion of cells expressing NKG7")
barplot(scmarkercluster,ylim = c(0,0.55),las=2,main='Seurat+SCMarker
',xlab="Cluster",ylab="The proportion of cells expressing NKG7")
```



FindMarker for each cluster

```
cluster=unique(Idents(GBMscmarker))
cluster=as.numeric(as.character(cluster[order(cluster)]))
gene=c()
for (i in cluster){
   cluster1.markers <- FindMarkers(GBMscmarker, only.pos = TRUE,ident
   .1 = i, min.pct = 0.25)
   if (length(grep("ERCC-",row.names(cluster1.markers)))!=0){
     cluster1.markers=cluster1.markers[-grep("ERCC-",row.names(cluster1.markers)),]
   }
   cluster1.markers=cluster1.markers[order(cluster1.markers$p_val_adj),]
   gene=union(gene[!is.na(gene)],row.names(cluster1.markers)[1:6])
}</pre>
```

"NKG7" %in% gene

```
## [1] TRUE
```

DoHeatmap(GBMscmarker, features = gene,size=3) + NoLegend()+ theme(a
xis.text.y = element\_text(size = 3))