## Single Cell Sequencing Analysis

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Dataset is downloaded from Broad Insitute Single cell website-
https://singlecell.broadinstitute.org/single_cell
/study/SCP1256/visium-demo-study#study-summary
Named as - Study: Visium demo study
Codes for single cell seq analysis
install.packages("Seurat")
install.packages("tidyverse")
install.packages("dplyr")
install.packages("patchwork")
install.packages("ggplot2")
library(Seurat)
library(tidyverse)
library(dplyr)
library(patchwork)
library(ggplot2)
##load the datasets##
da.data <- Read10X(data.dir ="C:/Users/Divya Agrawal/Downloads/")</pre>
da <- CreateSeuratObject(counts = da.data, min.cells = 4, min.features = 210)</pre>
##QC and filtering##
da[["percent.mt"]] <- PercentageFeatureSet(da, pattern = "^MT-")</pre>
plot1<-FeatureScatter(da, feature1 = "nCount_RNA", feature2 = "nFeature_RNA")</pre>
plot1
plot2 <-FeatureScatter(da, feature1 = "nCount_RNA", feature2 = "percent.mt")</pre>
plot2
plot1 +plot2
da <- subset(da, subset = nFeature_RNA > 215 & nFeature_RNA > 2500 & percent.mt <5)
##normalise the data##
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da <- NormalizeData(da, normalization.method = "LogNormalize", scale.factor = 10000)
##Find variable Features##
da <-FindVariableFeatures(da, selection.method = "vst", mfeatures=2000)</pre>
tp10<- head(VariableFeatures(da), 10)</pre>
tp10
plot1<- VariableFeaturePlot(da)</pre>
##scale the data##
all.genes <-rownames(da)
pre_scaling <-da</pre>
da <- ScaleData(da, features = all.genes)</pre>
##run linear dimensionality reduction ##
da<-RunPCA(da, features = VariableFeatures(object = da))</pre>
print(da[["pca"]], dims = 1:5, nfeatures = 5)
VizDimLoadings(da, dims = 1:2, reduction= "pca")
DimHeatmap(da, dims = 1, cells = 500)
DimHeatmap(da, dims = 1:15, cells = 500)
da <- JackStraw(da, num.replicate = 100)</pre>
JackStrawPlot(da, dims = 1:20)
da <-ScoreJackStraw(da, dims = 1:20)</pre>
da <-ScoreJackStraw(da, dims = 1:20)</pre>
JackStrawPlot(da, dims = 1:20)
##cluster##
da <- FindNeighbors(da, dims = 1:10)</pre>
da <-FindClusters(da, resolution = 0.5)</pre>
head(Idents(da),5)
##run non linear dimensionality reduction on top of dimensionality reduction
da <- RunUMAP(da, dims = 1:10)</pre>
DimPlot(da, reduction = "umap")
##assign the biological meaning to these clusters
da.markers <-FindAllMarkers(da, only.pos = TRUE, min.pct = 0.25, logfc.threshold = 0.25)
da.markers %>% group_by(cluster) %>% slice_max(n=2, order_by =avg_log2FC)
da.markers
```

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FeaturePlot(da, features = c("CTSH","CCL5","ENG","CD79A"))

##talk to a biologist

new.cluster.ids <- c("Naive CD T", "CD14+ Mono", "Memory CD4 T", "B","CD 8 T",
   "FCG3A+ Mono","NK cells", "DC" ,"platelet", "MAC complex")
names(new.cluster.ids) <- levels(da)
da <- RenameIdents(da, new.cluster.ids)
DimPlot(da, reduction = "umap", label = TRUE, pt.size = 0.5) + NoLegend()</pre>
```

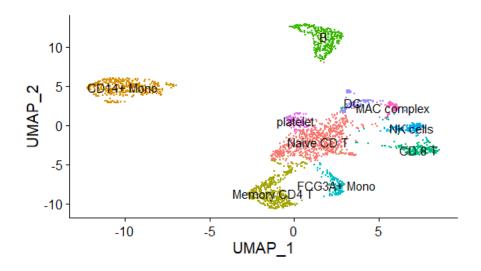


Figure 1: UMAP Plot