

## Visualizing MERFISH data using VeloViz

In this example, we will use VeloViz to create a 2D embedding to visualize MERFISH data collected from U-2 OS cells in culture. Since this data comes from a cell line in culture, we expect the main temporal signal to be progression through the cell-cycle. We will compare the VeloViz embedding to other commonly used embeddings. We will also compare the results we get when we restrict the input genes.

### Load preprocessed data

MERFISH data from Xia et. al., *PNAS*, 2019. This data is provided with the VeloViz package.

```
col = MERFISH$col
pcs = MERFISH$pcs
vel = MERFISH$vel

curr = vel$current
proj = vel$projected
```

### Load cell cycle genes

We will compare the VeloViz embeddings constructed with all genes to embeddings created using cell-cycle genes in the GO mitotic cell-cycle gene set and to embeddings created using genes that were found to have cell-cycle dependent expression in Xia et. al., *PNAS*, 2019.

```
merfish.genes = rownames(curr)

#GO cell cycle genes (GO:0000278)
# https://www.gsea-msigdb.org/gsea/msigdb/cards/GO_MITOTIC_CELL_CYCLE
cycle.genes.go = read.csv("GO_0000278.csv",header = FALSE)
cycle.genes.go = cycle.genes.go$V1

merfish.cycle.go = merfish.genes[which(merfish.genes %in% cycle.genes.go)]
curr.go = curr[merfish.cycle.go,]
proj.go = proj[merfish.cycle.go,]

#MERFISH genes exhibiting cell-cycle-dependent expression (Xia et al 2019, Supp Dataset 8)
# https://www.pnas.org/content/116/39/19490
cycle.genes.pnas = read.csv("pnas_sd08.csv",header = TRUE)
cycle.genes.pnas = cycle.genes.pnas$Gene

merfish.cycle.pnas = merfish.genes[which(merfish.genes %in% cycle.genes.pnas)]
curr.pnas = curr[merfish.cycle.pnas,]
proj.pnas = proj[merfish.cycle.pnas,]
```

## Build VeloViz embedding using all genes

```
par(mfrow = c(2,2))

veloviz.all = buildVeloviz(
  curr = curr,
  proj = proj,
  normalize.depth = TRUE,
  use.ods.genes = FALSE,
  pca = TRUE,
  nPCs = 3,
  center = TRUE,
  scale = TRUE,
  k = 100,
  similarity.threshold = 0,
  distance.weight = 1,
  distance.threshold = 1,
  weighted = TRUE,
  seed = 0,
  verbose = FALSE
)

## Warning in if (!class(curr) %in% c("dgCMatrx", "dgTMatrx")) {: the condition
## has length > 1 and only the first element will be used

## Warning in if (!class(proj) %in% c("dgCMatrx", "dgTMatrx")) {: the condition
## has length > 1 and only the first element will be used

emb.all.vv = veloviz.all$fdg_coords
plotEmbedding(emb.all.vv, colors = col[rownames(emb.all.vv)], main = 'all genes - veloviz')

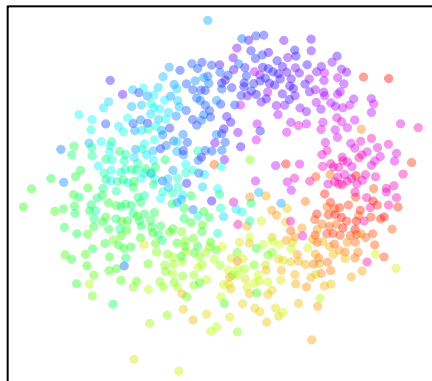
#PCA
emb.all.pca = pcs[,1:2]
plotEmbedding(emb.all.pca, colors = col, main = 'all genes - pca')

#tSNE
set.seed(0)
emb.all.tsne = Rtsne::Rtsne(pcs[,1:5], perplexity = 100)$Y
rownames(emb.all.tsne) = rownames(pcs)
plotEmbedding(emb.all.tsne, colors = col, main = 'all genes - t-SNE',
  xlab = "t-SNE X", ylab = "t-SNE y")

#UMAP
set.seed(0)
emb.all.umap = umap::umap(pcs[,1:5], min_dist = 0.3)$layout
rownames(emb.all.umap) = rownames(pcs)
plotEmbedding(emb.all.umap, colors = col, main = 'all genes - UMAP',
  xlab = "UMAP X", ylab = "UMAP Y")
```

**all genes – veloviz**

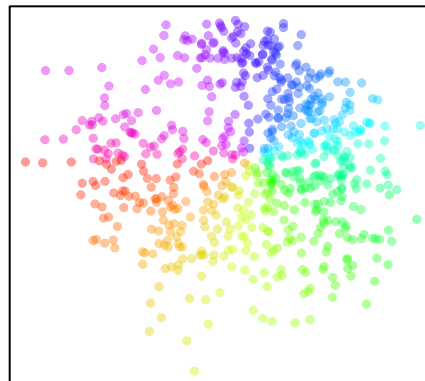
C2



C1

**all genes – pca**

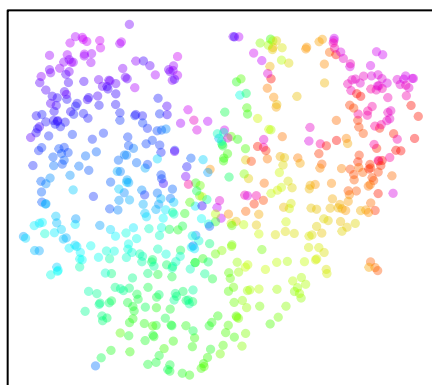
PC2



PC1

**all genes – t-SNE**

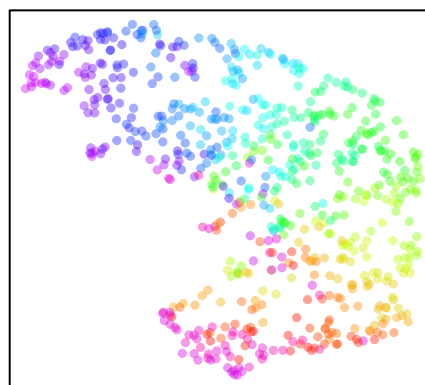
t-SNE y



t-SNE X

**all genes – UMAP**

UMAP Y



UMAP X

## Build VeloViz embedding using GO cell cycle genes

First, reduce dimensions.

```
curr.go.norm = normalizeDepth(curr.go)

## Warning in if (!class(counts) %in% c("dgCMatrx", "dgTMatrx")) {: the condition
## has length > 1 and only the first element will be used

## Converting to sparse matrix ...

## Normalizing matrix with 645 cells and 360 genes

curr.go.norm = normalizeVariance(curr.go.norm, details = TRUE)

## Using general additive modeling with k = 5...

## Identified 78 overdispersed genes using
##   adjusted p-value threshold alpha = 0.05

curr.go.norm = log10(curr.go.norm$matnorm + 1)
curr.go.pca = RSpectra::svds(A = t(as.matrix(curr.go.norm)), k = 50,
                             opts = list(center = TRUE, scale = FALSE,
                                           maxitr = 2000, tol = 1e-10))

curr.go.pca = curr.go.pca$u
rownames(curr.go.pca) = rownames(pcs)
```

Now, build embeddings.

```
par(mfrow = c(2,2))

veloviz.go = buildVeloviz(
  curr = curr.go,
  proj = proj.go,
  normalize.depth = TRUE,
  use.ods.genes = FALSE,
  pca = TRUE,
  nPCs = 3,
  center = TRUE,
  scale = TRUE,
  k = 20,
  similarity.threshold = 0,
  distance.weight = 0.1,
  distance.threshold = 1,
  weighted = TRUE,
  seed = 0,
  verbose = FALSE
)

## Warning in if (!class(curr) %in% c("dgCMatrx", "dgTMatrx")) {: the condition
## has length > 1 and only the first element will be used

## Warning in if (!class(proj) %in% c("dgCMatrx", "dgTMatrx")) {: the condition
## has length > 1 and only the first element will be used

emb.go.vv = veloviz.go$fdg_coords
plotEmbedding(emb.go.vv, colors = col[rownames(emb.go.vv)],
              main = 'GO cell cycle genes - veloviz')

#PCA
```

```

emb.go.pca = curr.go.pca[,1:2]
plotEmbedding(emb.go.pca, colors = col, main = 'GO cell cycle genes - pca')

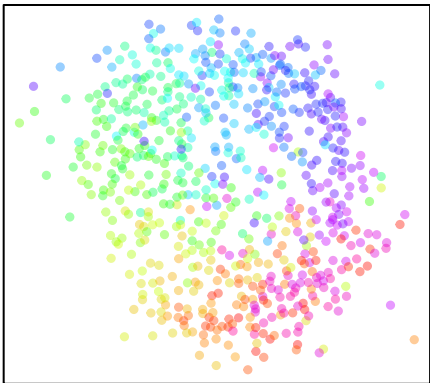
#tSNE
set.seed(0)
emb.go.tsne = Rtsne::Rtsne(curr.go.pca[,1:5], perplexity = 100)$Y
rownames(emb.go.tsne) = rownames(curr.go.pca)
plotEmbedding(emb.go.tsne, colors = col, main = 'GO cell cycle genes - t-SNE',
              xlab = "t-SNE X", ylab = "t-SNE y")

#UMAP
set.seed(0)
emb.go.umap = umap::umap(curr.go.pca[,1:5], min_dist = 0.3)$layout
rownames(emb.go.umap) = rownames(curr.go.pca)
plotEmbedding(emb.go.umap, colors = col, main = 'GO cell cycle genes - UMAP',
              xlab = "UMAP X", ylab = "UMAP Y")

```

GO cell cycle genes – veloviz

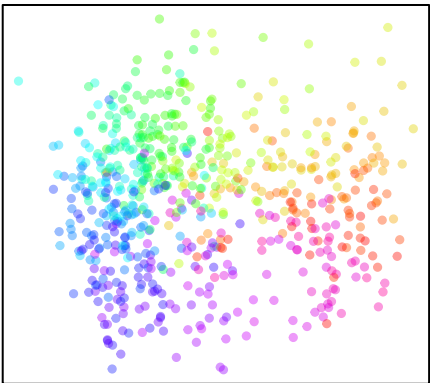
C2



C1

GO cell cycle genes – pca

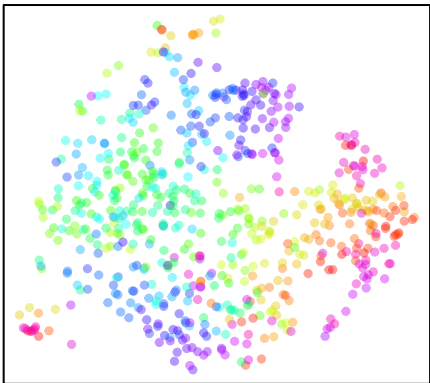
emb[,2]



emb[,1]

GO cell cycle genes – t-SNE

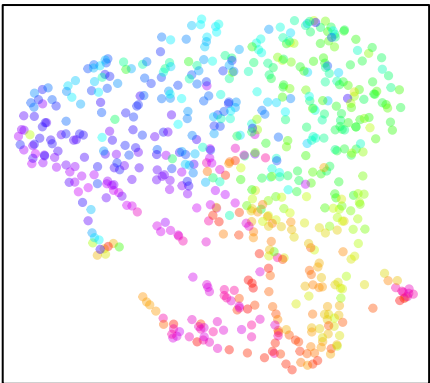
t-SNE y



t-SNE X

GO cell cycle genes – UMAP

UMAP Y



UMAP X

## Build VeloViz embedding with cell-cycle dependent genes

First, reduce dimensions.

```
curr.pnas.norm = normalizeDepth(curr.pnas)

## Warning in if (!class(counts) %in% c("dgCMatrx", "dgTMatrx")) {: the condition
## has length > 1 and only the first element will be used

## Converting to sparse matrix ...

## Normalizing matrix with 645 cells and 1471 genes

curr.pnas.norm = normalizeVariance(curr.pnas.norm, details = TRUE)

## Using general additive modeling with k = 5...

## Identified 389 overdispersed genes using
##   adjusted p-value threshold alpha = 0.05

curr.pnas.norm = log10(curr.pnas.norm$matnorm + 1)
curr.pnas.pca = RSpectra::svds(A = t(as.matrix(curr.pnas.norm)), k = 50,
                             opts = list(center = TRUE, scale = FALSE,
                                           maxitr = 2000, tol = 1e-10))

curr.pnas.pca = curr.pnas.pca$u
rownames(curr.pnas.pca) = rownames(pcs)
```

Now, build embeddings.

```
par(mfrow = c(2,2))

veloviz.pnas = buildVeloviz(
  curr = curr.pnas,
  proj = proj.pnas,
  normalize.depth = TRUE,
  use.ods.genes = FALSE,
  pca = TRUE,
  nPCs = 3,
  center = TRUE,
  scale = TRUE,
  k = 50,
  similarity.threshold = 0.5,
  distance.weight = 0.01,
  distance.threshold = 1,
  weighted = TRUE,
  seed = 0,
  verbose = FALSE
)

## Warning in if (!class(curr) %in% c("dgCMatrx", "dgTMatrx")) {: the condition
## has length > 1 and only the first element will be used

## Warning in if (!class(proj) %in% c("dgCMatrx", "dgTMatrx")) {: the condition
## has length > 1 and only the first element will be used

emb.pnas.vv = veloviz.pnas$fdg_coords
plotEmbedding(emb.pnas.vv, colors = col[rownames(emb.pnas.vv)],
              main = 'Xia et al cell cycle genes - veloviz')

#PCA
```

```

emb.pnas.pca = curr.pnas.pca[,1:2]
plotEmbedding(emb.pnas.pca, colors = col, main = 'Xia et al cell cycle genes - pca')

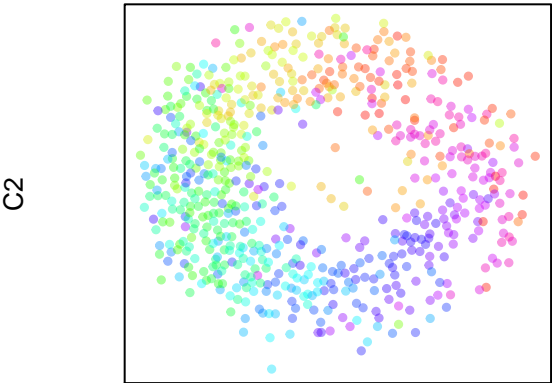
#tSNE
set.seed(0)
emb.pnas.tsne = Rtsne::Rtsne(curr.pnas.pca[,1:5], perplexity = 100)$Y
rownames(emb.pnas.tsne) = rownames(curr.pnas.pca)
plotEmbedding(emb.pnas.tsne, colors = col, main = 'Xia et al cell cycle genes - t-SNE',
              xlab = "t-SNE X", ylab = "t-SNE y")

#UMAP
set.seed(0)
emb.pnas.umap = umap::umap(curr.pnas.pca[,1:5], min_dist = 0.3)$layout
rownames(emb.pnas.umap) = rownames(curr.pnas.pca)
plotEmbedding(emb.pnas.umap, colors = col, main = 'Xia et al cell cycle genes - UMAP',
              xlab = "UMAP X", ylab = "UMAP Y")

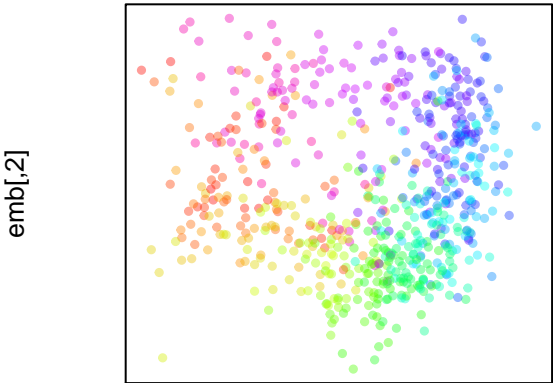
```



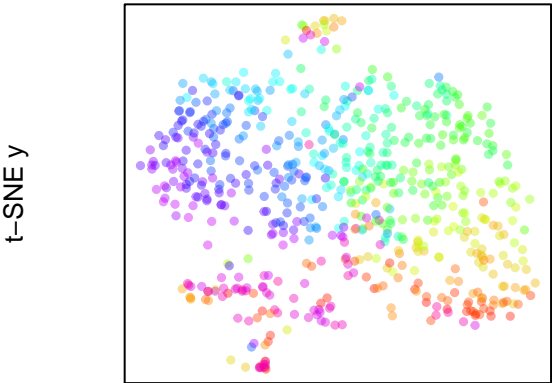
**Xia et al cell cycle genes – veloviz**



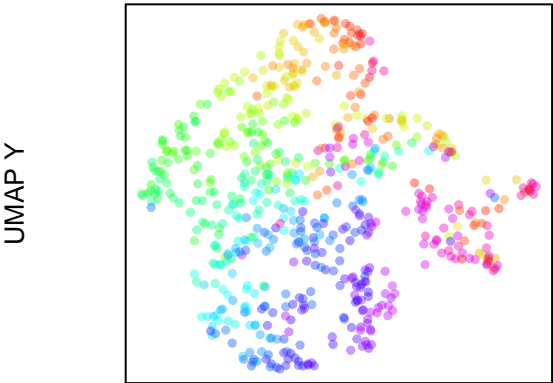
**Xia et al cell cycle genes – pca**



**Xia et al cell cycle genes – t-SNE**



**Xia et al cell cycle genes – UMAP**



## Comparing VeloViz embeddings

```
par(mfrow = c(1,3))

plotEmbedding(emb.all.vv, colors = col[rownames(emb.pnas.vv)],
              main = 'all genes - veloviz')

plotEmbedding(emb.go.vv, colors = col[rownames(emb.pnas.vv)],
              main = 'GO cell cycle genes - veloviz')

plotEmbedding(emb.pnas.vv, colors = col[rownames(emb.pnas.vv)],
              main = 'Xia et al cell cycle genes - veloviz')
```

