

quollr: An R Package for Visualizing 2D Models in High Dimensional Space

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Abstract An abstract of less than 150 words.

```
#library(quollr)
library(readr)
library(ggplot2)
library(dplyr)
library(ggbeeswarm)
library(Rtsne)
library(umap)
library(phateR)
library(reticulate)
library(rsample)

set.seed(20230531)

use_python("~/miniforge3/envs/pcamp_env/bin/python")
use_condaenv("pcamp_env")

reticulate::source_python(paste0(here::here(), "/scripts/function_scripts/Fit_PacMAP_code.py"))
reticulate::source_python(paste0(here::here(), "/scripts/function_scripts/Fit_TriMAP_code.py"))
```

1 Introduction

2 Methodology

Usage

- dependancies

```
library(tools)
package_dependencies("quollr")

• basic example ->
```

Compute hexagonal bin configurations

```
num_bins_x <- calculate_effective_x_bins(.data = s_curve_noise_umap, x = "UMAP1", hex_size = NA)
num_bins_x

#> [1] 4

num_bins_y <- calculate_effective_y_bins(.data = s_curve_noise_umap, y = "UMAP2", hex_size = NA)
num_bins_y

#> [1] 8
```

Generate full hex grid

Generating full hexagonal grid contains main three steps:

1. Generate all the hexagonal bin centroids

Steps:

- First compute hex grid bound values along the x and y axis and generate all the points within the hex box

```
cell_area <- 1

hex_size <- sqrt(2 * cell_area / sqrt(3))

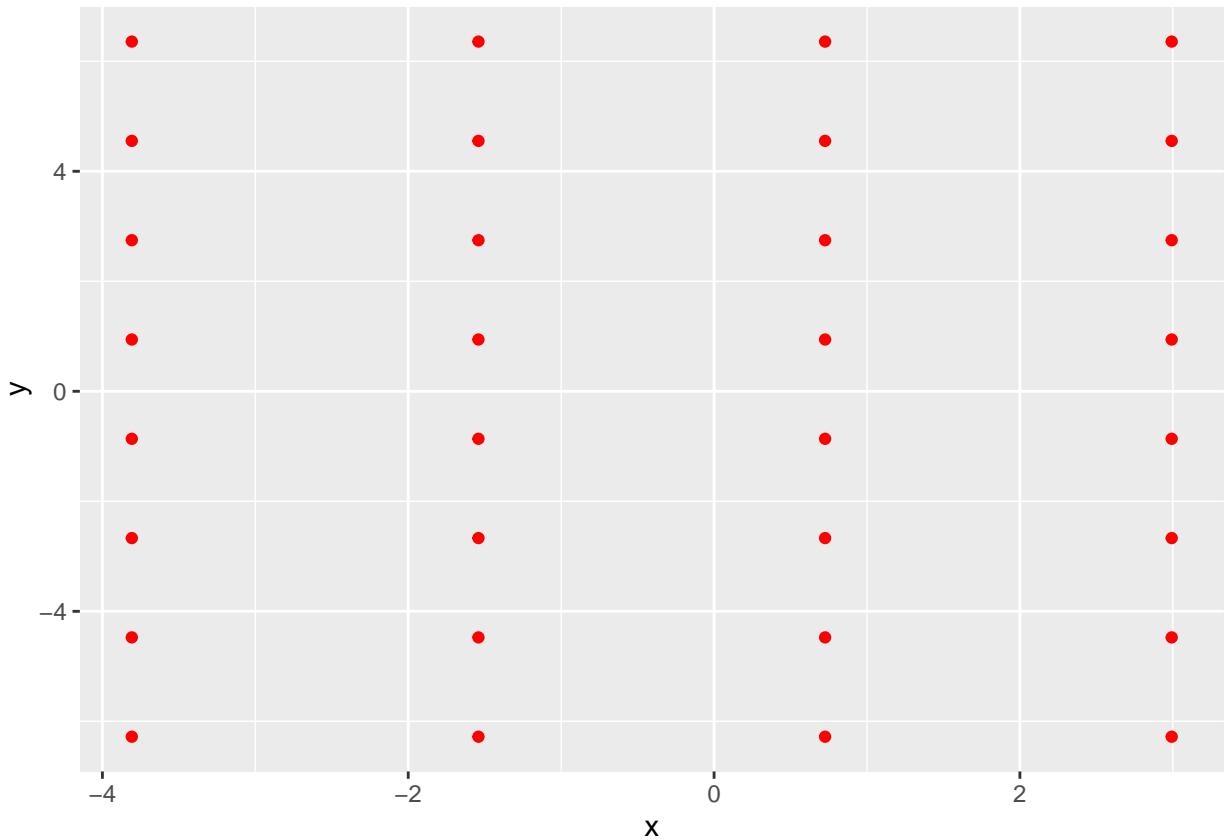
buffer_size <- hex_size/2

x_bounds <- seq(min(s_curve_noise_umap[["UMAP1"]]) - buffer_size,
                 max(s_curve_noise_umap[["UMAP1"]]) + buffer_size, length.out = num_bins_x)

y_bounds <- seq(min(s_curve_noise_umap[["UMAP2"]]) - buffer_size,
                 max(s_curve_noise_umap[["UMAP2"]]) + buffer_size, length.out = num_bins_y)

box_points <- expand.grid(x = x_bounds, y = y_bounds)

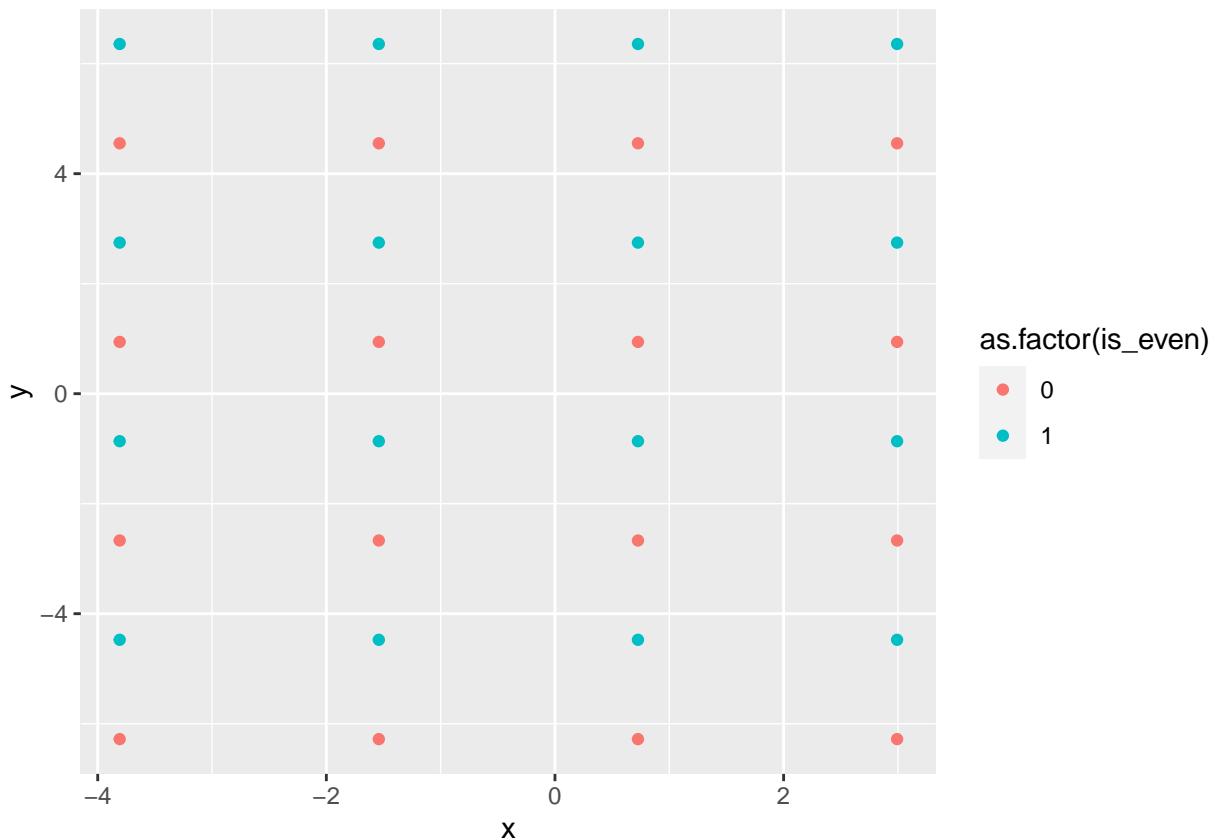
ggplot() +
  geom_point(data = box_points, aes(x = x, y = y), color = "red")
```



- Second for each x-value, find which y values are in the even row

```
box_points <- box_points |>
  dplyr::arrange(x) |>
  dplyr::group_by(x) |>
  dplyr::group_modify(~ generate_even_y(.x)) |>
  tibble::as_tibble()

ggplot() +
  geom_point(data = box_points,
             aes(x = x, y = y, colour = as.factor(is_even)))
```



- Then, shift the x values of the even rows

```
## Shift for even values in x-axis
x_shift <- unique(box_points$x)[2] - unique(box_points$x)[1]

box_points$x <- box_points$x + x_shift/2 * ifelse(box_points$is_even == 1, 1, 0)

ggplot() +
  geom_point(data = box_points, aes(x = x, y = y), color = "red")
```

```
glimpse(all_centroids_df)
```

2. Generate hexagonal coordinates

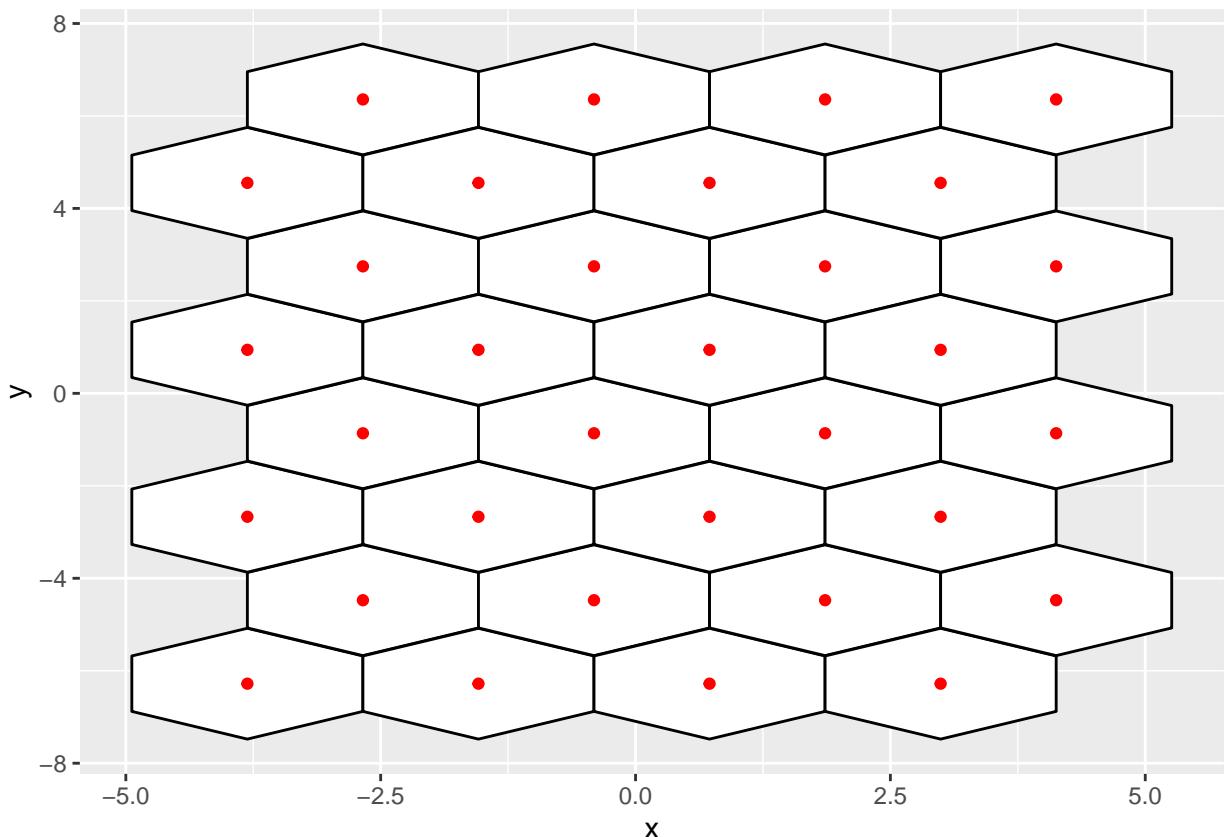
Steps: - Compute horizontal width of the hexagon

- Compute vertical width of the hexagon and multiply by a factor for overlapping ($\sqrt{3}/2 * 1.15$)
 - Obtain hexagon polygon coordinates
 - Obtain the number of hexagons in the full grid
 - Generate the coordinates for the hexagons

```
hex_grid <- gen_hex_coordinates(all_centroids_df, hex_size = NA)
glimpse(hex_grid)
```

```
#> Rows: 192  
#> Columns: 3  
#> $ x <dbl> -2.674222, -2.674222, -3.807643, -4.941063, -4.941063, -3.807643, ~  
#> $ y <dbl> -5.6804828, -6.8791681, -7.4785108, -6.8791681, -5.6804828, -5.0811~  
#> $ id <int> 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 3, 3, 3, 3, 3, 3, 4, 4, 4, 4~
```

```
ggplot(data = hex_grid, aes(x = x, y = y)) + geom_polygon(fill = "white", color = "black", aes(group = id)) +
  geom_point(data = all_centroids_df, aes(x = x, y = y), color = "red")
```



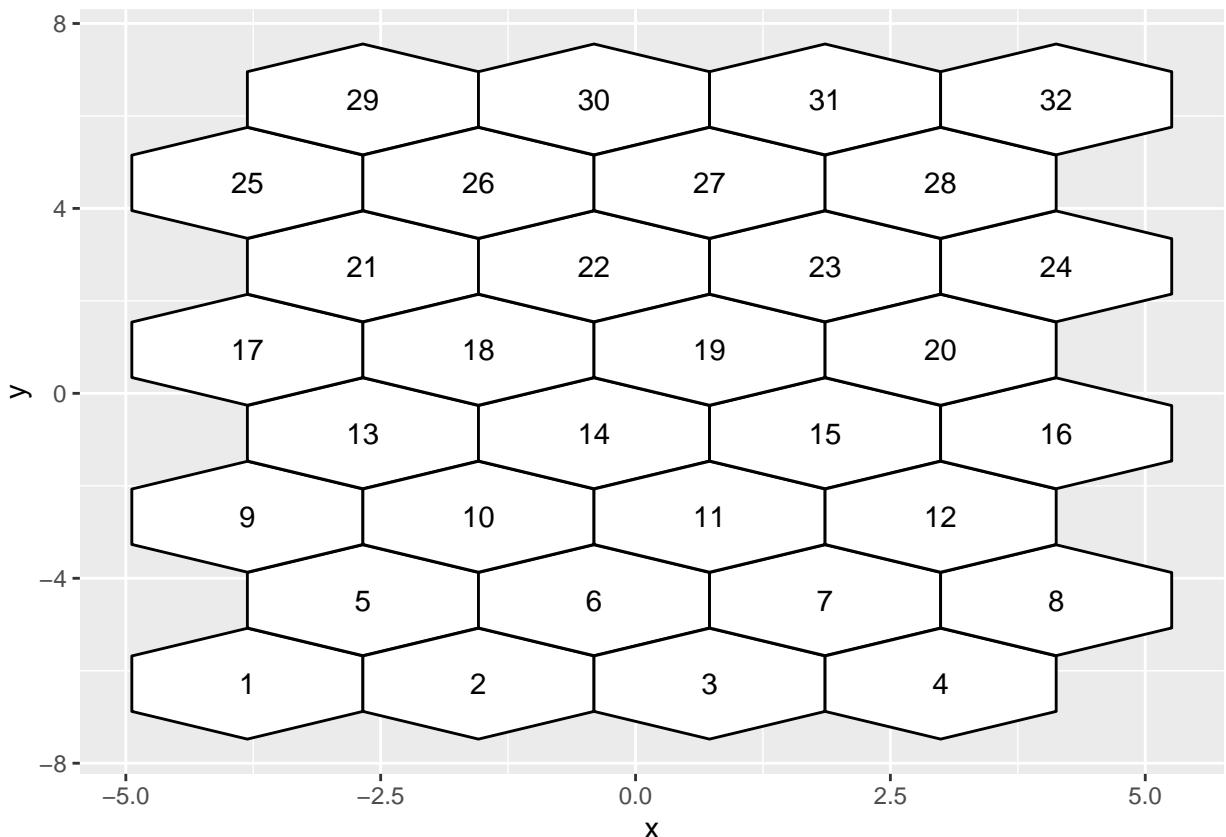
3. Map hexagonal IDs

Steps:

- Filter the data set with specific y value
- Order the x values for a specific y value
- Repeat the process for all unique y values

```
full_grid_with_hexbin_id <- map_hexbin_id(all_centroids_df)
```

```
ggplot(data = hex_grid, aes(x = x, y = y)) + geom_polygon(fill = "white", color = "black", aes(group = id)) +  
  geom_text(data = full_grid_with_hexbin_id, aes(x = c_x, y = c_y, label = hexID))
```



4. Map polygon IDs

Steps:

- Filter specific hexagon
- Filter specific polygon
- Check the selected hexagonal centroid exists within the polygon
- if so assign that id to centroid, if not check until find the polygon which contains the centroid

```
full_grid_with_polygon_id <- map_polygon_id(full_grid_with_hexbin_id, hex_grid)
```

4. Assign data into hexagons

- Compute distances between nldr coordinates and hex bin centroids
- Find the hexagonal centroid that have the minimum distance

```
s_curve_noise_umap_with_id <- assign_data(s_curve_noise_umap, full_grid_with_hexbin_id)
```

5. Compute standardized counts

- Compute number of data points within each hexagon
- Compute standardise count by dividing the counts by the maximum

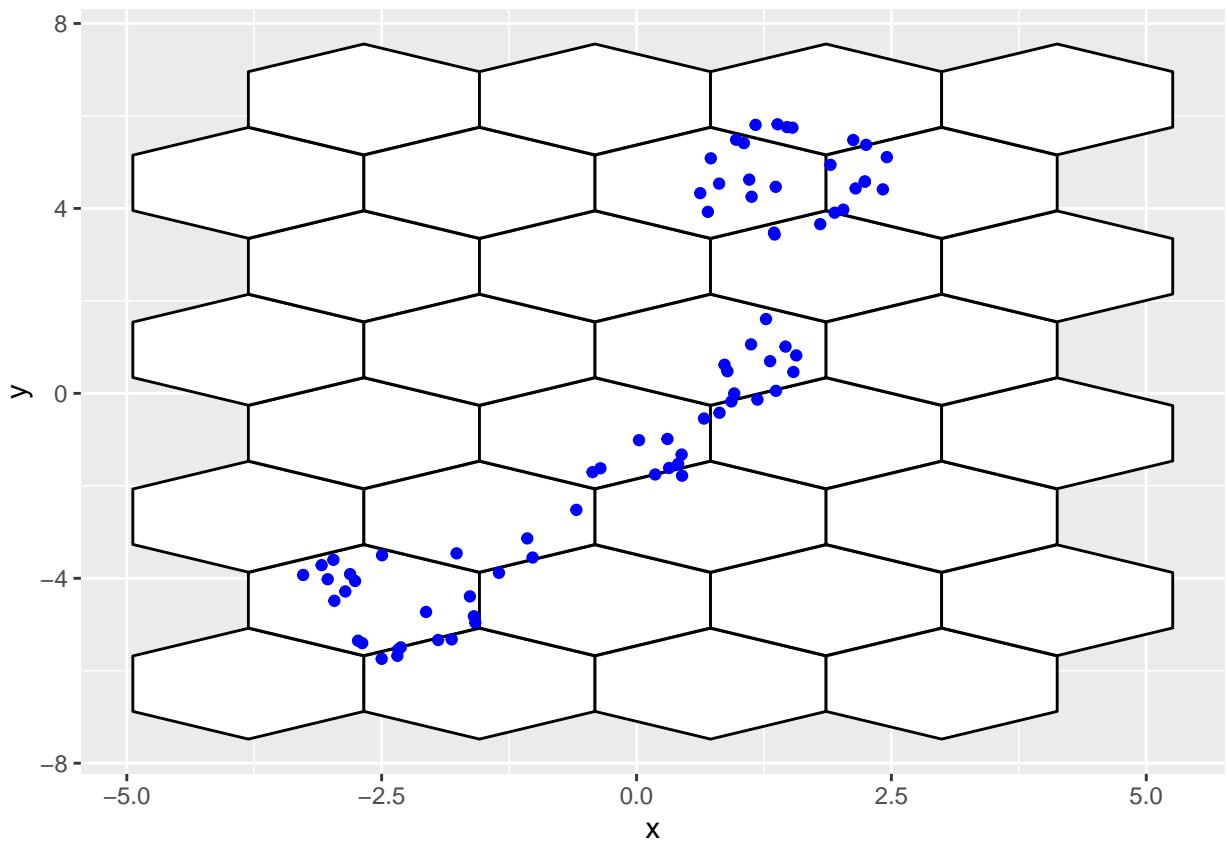
```
df_with_std_counts <- compute_std_counts(nldr_df = s_curve_noise_umap_with_id)
```

6. Extract full grid info

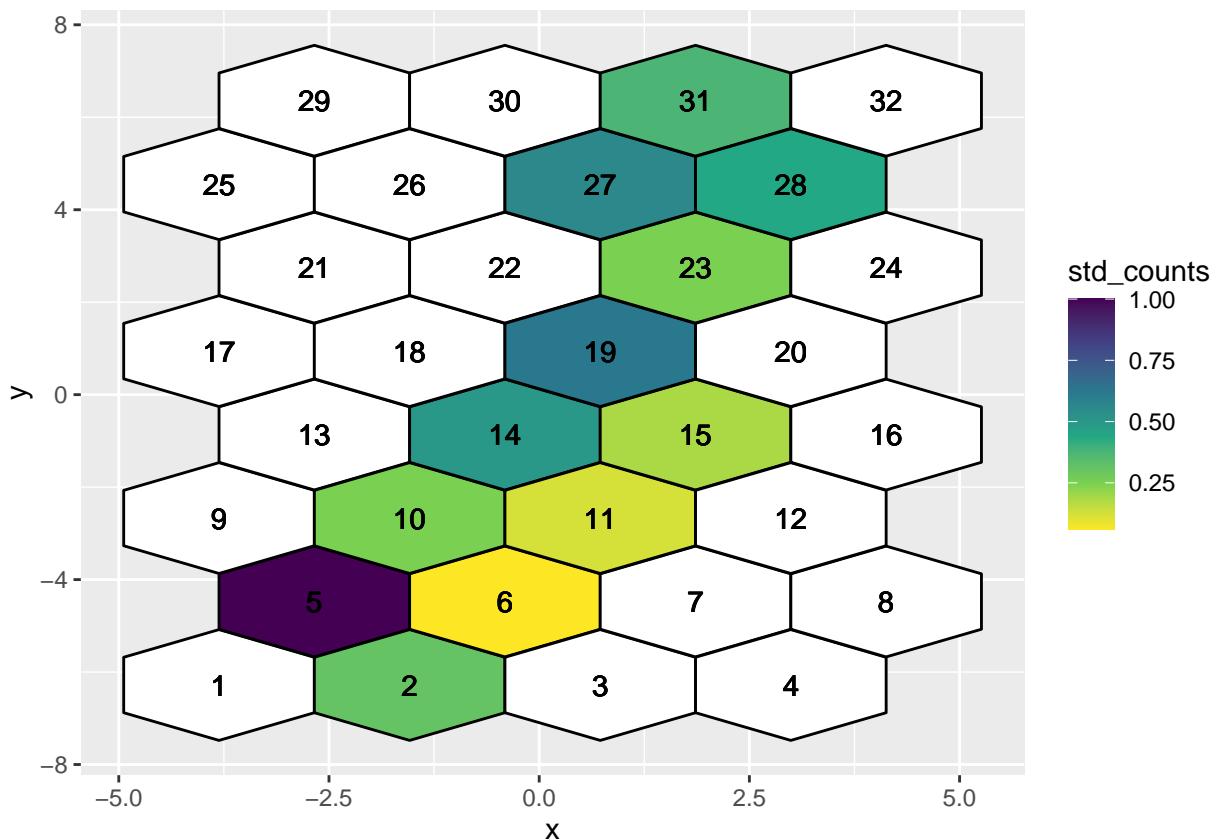
- Assign standardize counts for hex bins
- Join with the hexagonal coordinates

```
hex_full_count_df <- generate_full_grid_info(full_grid_with_polygon_id, df_with_std_counts, hex_grid)
```

```
ggplot(data = hex_grid, aes(x = x, y = y)) + geom_polygon(fill = "white", color = "black", aes(group = id)) +
  geom_point(data = s_curve_noise_umap, aes(x = UMAP1, y = UMAP2), color = "blue")
```



```
ggplot(data = hex_full_count_df, aes(x = x, y = y)) +
  geom_polygon(color = "black", aes(group = polygon_id, fill = std_counts)) +
  geom_text(aes(x = c_x, y = c_y, label = hexID)) +
  scale_fill_viridis_c(direction = -1, na.value = "#ffffff")
```



Buffer size When generating hexagonal bins in R, a buffer is often included to ensure that the data points are evenly distributed within the bins and to prevent edge effects. The buffer helps in two main ways:

1. **Preventing Edge Effects:** Without a buffer, the outermost data points might fall near the boundary of the hexagonal grid, leading to incomplete bins or uneven distribution of data. By adding a buffer, you create a margin around the outer edges of the grid, ensuring that all data points are fully enclosed within the bins.
2. **Ensuring Even Distribution:** The buffer allows for a smoother transition between adjacent bins. This helps in cases where data points are not perfectly aligned with the grid lines, ensuring that each data point is assigned to the nearest bin without bias towards any specific direction.

Overall, including a buffer when generating hexagonal bins helps to produce more accurate and robust binning results, particularly when dealing with real-world data that may have irregular distributions or boundary effects.

Construct the 2D model with different options

Construct the high-D model with different options

```
## To generate a data set with high-D and 2D training data
df_all <- training_data |> dplyr::select(-ID) |>
  dplyr::bind_cols(s_curve_noise_umap_with_id)

## To generate averaged high-D data

df_bin <- avg_highD_data(.data = df_all, column_start_text = "x") ## Need to pass ID column name
```

Generate the triangular mesh

```
df_bin_centroids <- hex_full_count_df[complete.cases(hex_full_count_df[["std_counts"]]), ] |>
  dplyr::select("c_x", "c_y", "hexID", "std_counts") |>
  dplyr::distinct() |>
  dplyr::rename(c("x" = "c_x", "y" = "c_y"))

df_bin_centroids

#>      x       y hexID std_counts
#> 1 -2.6742223 -4.4744481     5   1.0000
#> 2 -1.5408019 -6.2798254     2   0.3125
#> 3 -0.4073814 -4.4744481     6   0.0625
#> 4 -1.5408019 -2.6690708    10   0.2500
#> 5 -0.4073814 -0.8636935    14   0.5000
#> 6  0.7260390 -2.6690708    11   0.1250
#> 7  1.8594594 -0.8636935    15   0.1875
#> 8  0.7260390  0.9416838    19   0.6250
#> 9  1.8594594  2.7470611    23   0.2500
#> 10 0.7260390  4.5524384    27   0.5625
#> 11 1.8594594  6.3578158    31   0.3750
#> 12 2.9928798  4.5524384    28   0.4375

tr1_object <- triangulate_bin_centroids(df_bin_centroids, x, y)
tr_from_to_df <- generate_edge_info(triangular_object = tr1_object)
```

Compute parameter defaults

Shift the hexagonal grid origin If shift_x happen to the positive direction of x it should input as a positive value, if not other way If shift_y happen to the positive direction of y it should input as a positive value, if not other way

1. Assign shift along the x and y axis (limited the amount should less than the cell_diameter)
2. Generate bounds with shift origin

```

all_centroids_df_shift <- extract_coord_of_shifted_hex_grid(nldr_df = s_curve_noise_umap,
  x = "UMAP1", y = "UMAP2",
  num_bins_x = num_bins_x,
  num_bins_y = num_bins_y,
  shift_x = 0.2690002, shift_y = 0.271183,
  buffer_size = NA, hex_size = NA)

glimpse(all_centroids_df_shift)

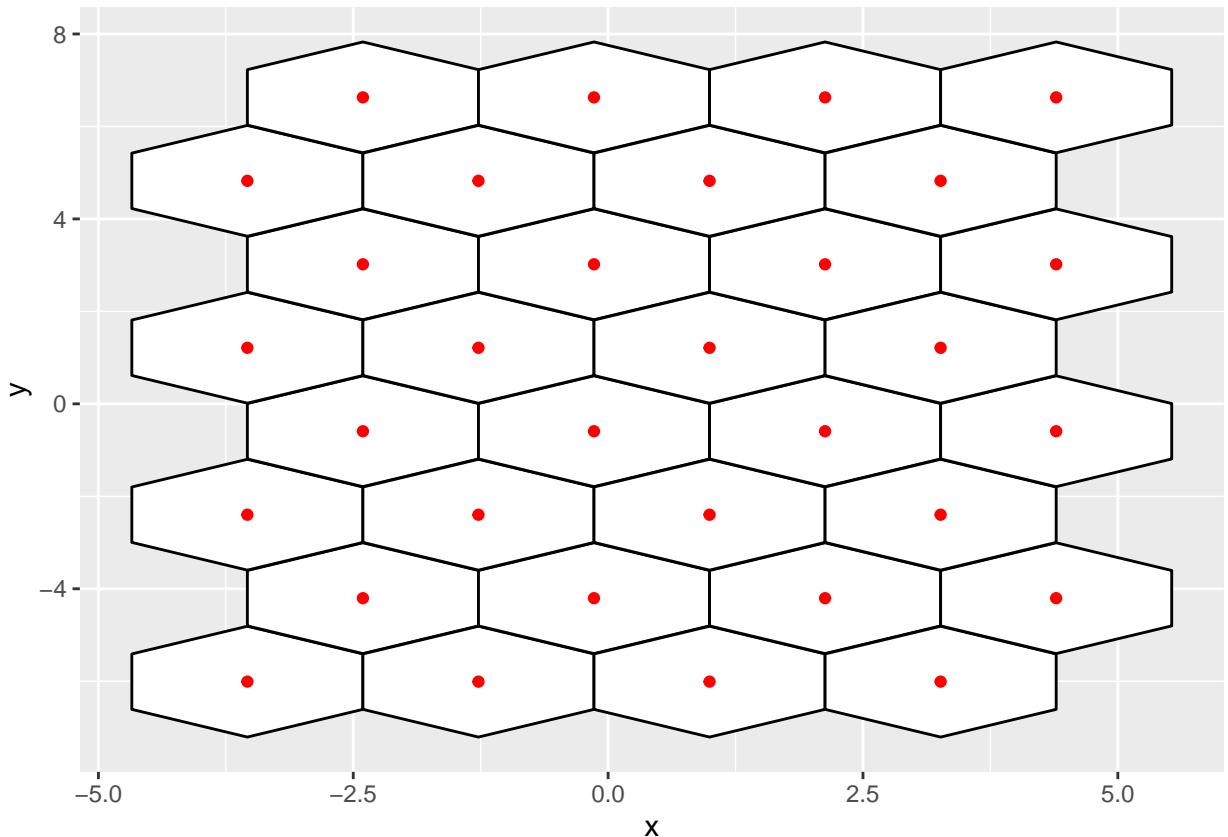
#> Rows: 32
#> Columns: 2
#> $ x <dbl> -3.5386425, -2.4052221, -3.5386425, -2.4052221, -3.5386425, -2.40522~
#> $ y <dbl> -6.0086424, -4.2032651, -2.3978878, -0.5925105, 1.2128668, 3.0182441~

hex_grid <- gen_hex_coordinates(all_centroids_df_shift)
glimpse(hex_grid)

#> Rows: 192
#> Columns: 3
#> $ x <dbl> -2.405222, -2.405222, -3.538643, -4.672063, -4.672063, -3.538643, ~
#> $ y <dbl> -5.409299776, -6.607985117, -7.207327787, -6.607985117, -5.40929977~
#> $ id <int> 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 3, 3, 3, 3, 3, 4, 4, 4, 4~

ggplot(data = hex_grid, aes(x = x, y = y)) + geom_polygon(fill = "white", color = "black", aes(group = id)) +
  geom_point(data = all_centroids_df_shift, aes(x = x, y = y), color = "red")

```

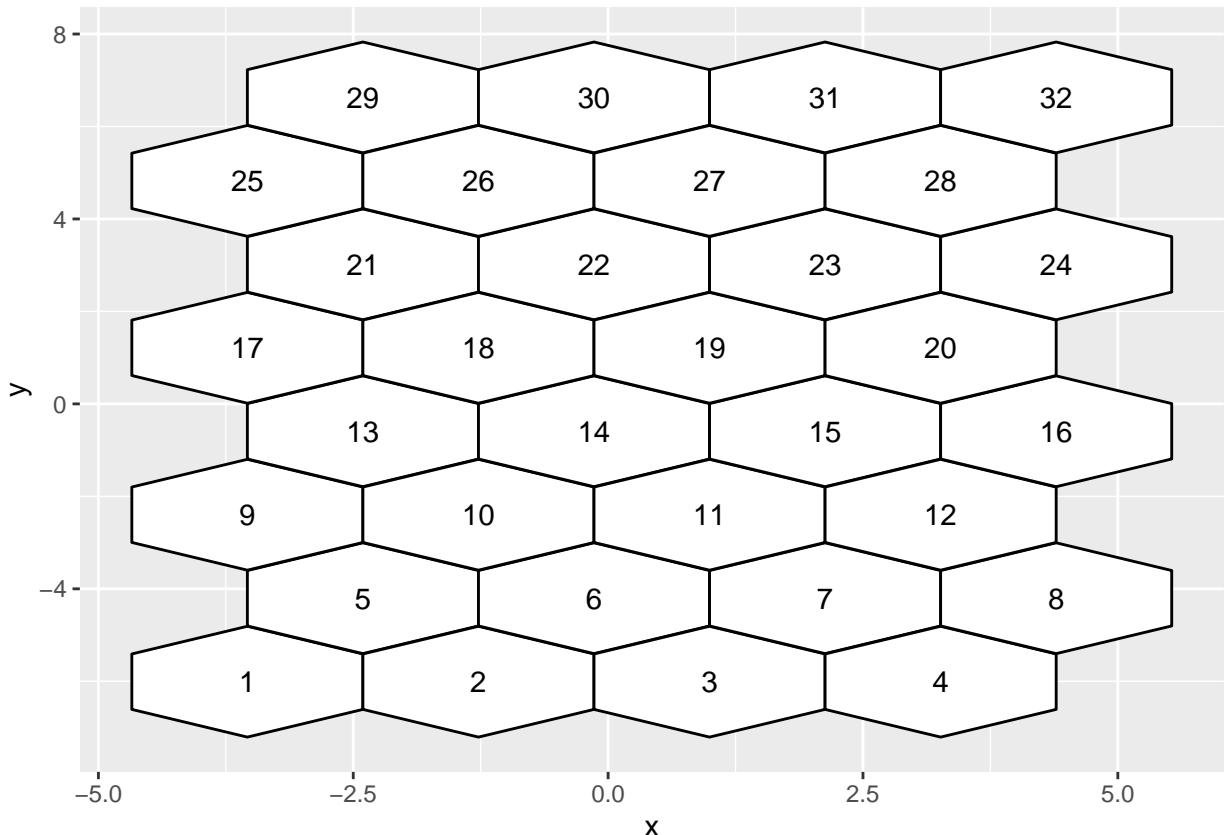


```

full_grid_with_hexbin_id <- map_hexbin_id(all_centroids_df_shift)

ggplot(data = hex_grid, aes(x = x, y = y)) + geom_polygon(fill = "white", color = "black", aes(group = id)) +
  geom_text(data = full_grid_with_hexbin_id, aes(x = c_x, y = c_y, label = hexID))

```



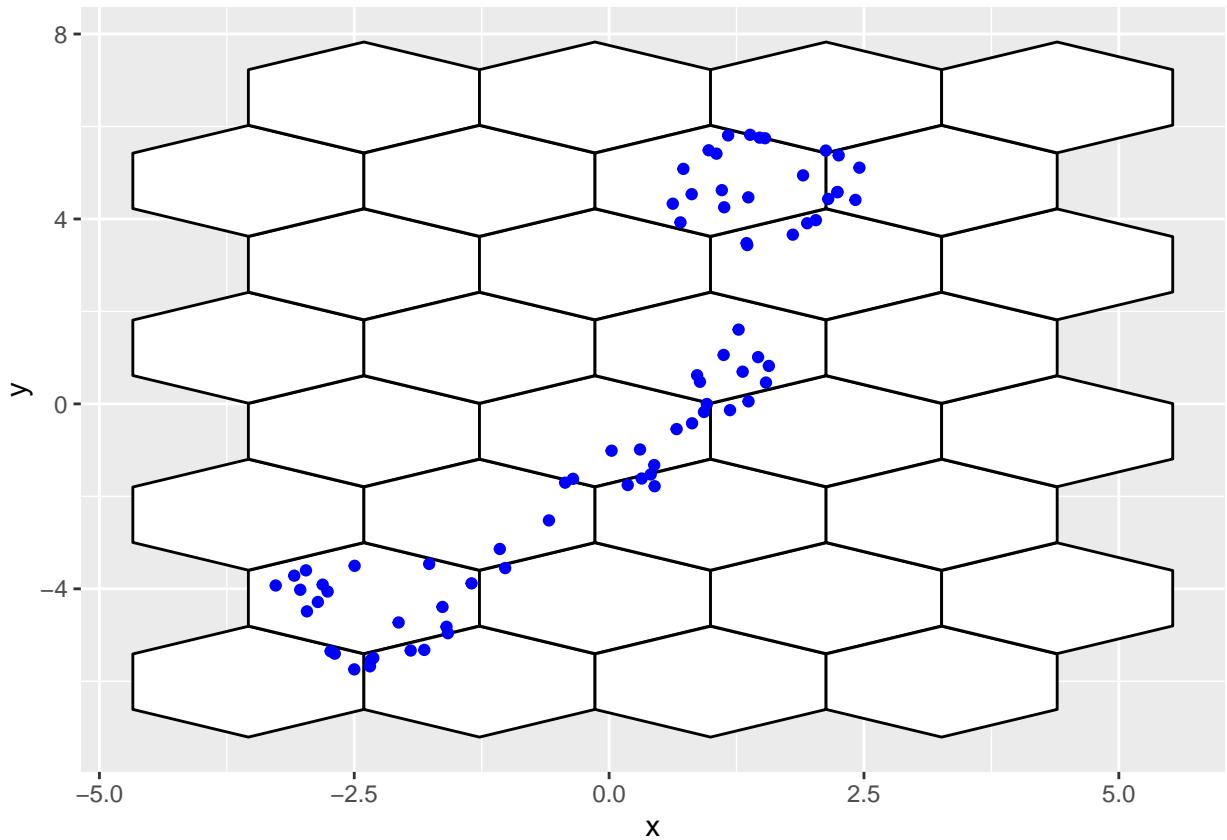
```
full_grid_with_polygon_id <- map_polygon_id(full_grid_with_hexbin_id, hex_grid)

s_curve_noise_umap_with_id <- assign_data(s_curve_noise_umap, full_grid_with_hexbin_id)

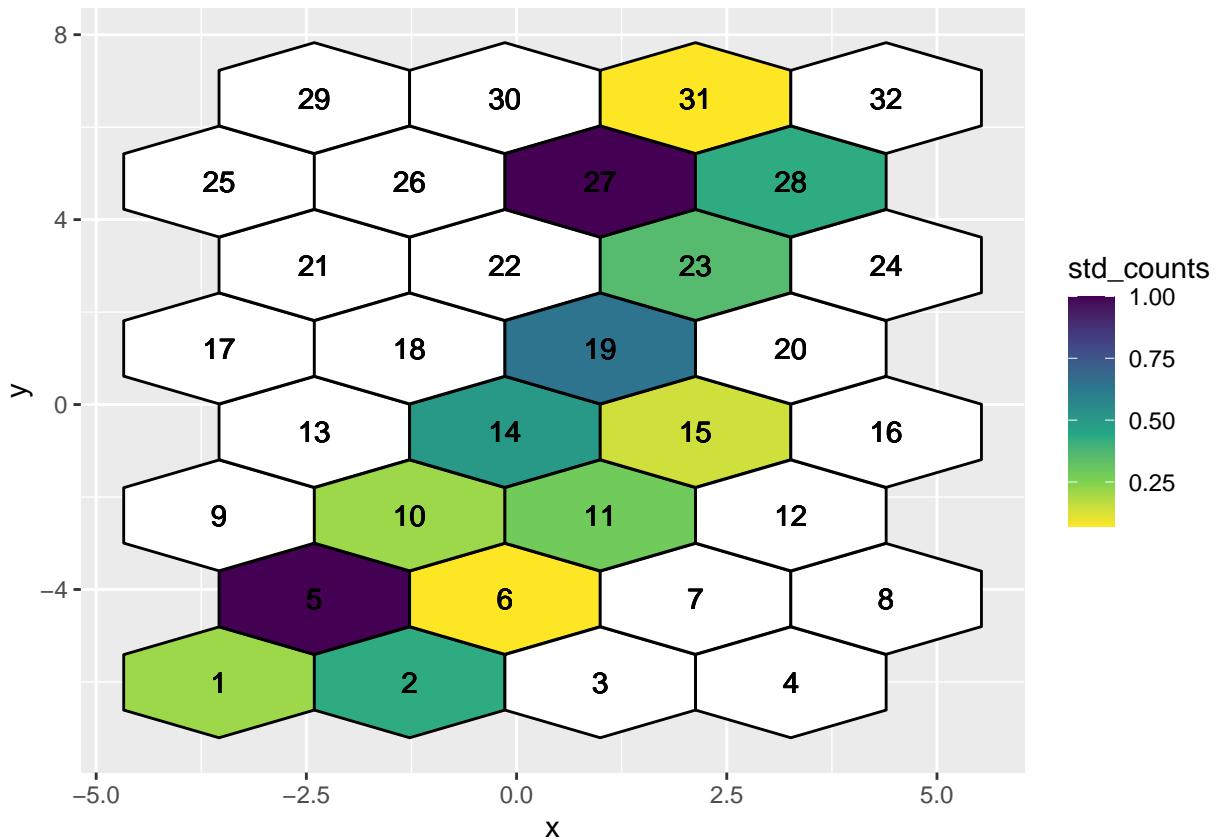
df_with_std_counts <- compute_std_counts(nldr_df = s_curve_noise_umap_with_id)

hex_full_count_df <- generate_full_grid_info(full_grid_with_polygon_id, df_with_std_counts, hex_grid)

ggplot(data = hex_grid, aes(x = x, y = y)) + geom_polygon(fill = "white", color = "black", aes(group = id)) +
  geom_point(data = s_curve_noise_umap, aes(x = UMAP1, y = UMAP2), color = "blue")
```



```
ggplot(data = hex_full_count_df, aes(x = x, y = y)) +
  geom_polygon(color = "black", aes(group = polygon_id, fill = std_counts)) +
  geom_text(aes(x = c_x, y = c_y, label = hexID)) +
  scale_fill_viridis_c(direction = -1, na.value = "#ffffff")
```



```
df_bin_centroids <- hex_full_count_df[complete.cases(hex_full_count_df[["std_counts"]]), ] |>
```

```

dplyr::select("c_x", "c_y", "hexID", "std_counts") |>
dplyr::distinct() |>
dplyr::rename(c("x" = "c_x", "y" = "c_y"))

df_bin_centroids

#>      x     y hexID std_counts
#> 1 -3.5386425 -6.0086424    1 0.21428571
#> 2 -2.4052221 -4.2032651    5 1.00000000
#> 3 -1.2718017 -6.0086424    2 0.42857143
#> 4 -0.1383812 -4.2032651    6 0.07142857
#> 5 -1.2718017 -2.3978878   10 0.21428571
#> 6 -0.1383812 -0.5925105   14 0.50000000
#> 7  0.9950392 -2.3978878   11 0.28571429
#> 8  2.1284596 -0.5925105   15 0.14285714
#> 9  0.9950392  1.2128668   19 0.64285714
#> 10 2.1284596  3.0182441   23 0.35714286
#> 11 0.9950392  4.8236214   27 1.00000000
#> 12 2.1284596  6.6289988   31 0.07142857
#> 13 3.2618800  4.8236214   28 0.42857143

tr1_object <- triangulate_bin_centroids(df_bin_centroids, x, y)
tr_from_to_df <- generate_edge_info(triangular_object = tr1_object)

bin_centroids_shift <- ggplot(data = hex_full_count_df, aes(x = c_x, y = c_y)) +
  geom_point(color = "#bdbdbd") +
  geom_point(data = shifted_hex_coord_df, aes(x = c_x, y = c_y), color = "#feb24c") +
  coord_cartesian(xlim = c(-5, 8), ylim = c(-10, 10)) +
  theme_void() +
  theme(legend.position="none", legend.direction="horizontal", plot.title = element_text(size = 7, hjust = 0.5, vjust = 0.5),
        axis.title.x = element_blank(), axis.title.y = element_blank(),
        axis.text.x = element_blank(), axis.ticks.x = element_blank(),
        axis.text.y = element_blank(), axis.ticks.y = element_blank(),
        panel.grid.major = element_blank(), panel.grid.minor = element_blank(), #change legend key width
        legend.title = element_text(size=8), #change legend title font size
        legend.text = element_text(size=6)) +
  guides(fill = guide_colourbar(title = "Standardized count")) +
  annotate(geom = 'text', label = "a", x = -Inf, y = Inf, hjust = -0.3, vjust = 1, size = 3)

hex_grid_shift <- ggplot(data = shifted_hex_coord_df, aes(x = x, y = y)) +
  geom_polygon(fill = NA, color = "#feb24c", aes(group = polygon_id)) +
  geom_polygon(data = hex_full_count_df, aes(x = x, y = y, group = polygon_id),
               fill = NA, color = "#bdbdbd") +
  coord_cartesian(xlim = c(-5, 8), ylim = c(-10, 10)) +
  theme_void() +
  theme(legend.position="none", legend.direction="horizontal", plot.title = element_text(size = 7, hjust = 0.5, vjust = 0.5),
        axis.title.x = element_blank(), axis.title.y = element_blank(),
        axis.text.x = element_blank(), axis.ticks.x = element_blank(),
        axis.text.y = element_blank(), axis.ticks.y = element_blank(),
        panel.grid.major = element_blank(), panel.grid.minor = element_blank(), #change legend key width
        legend.title = element_text(size=8), #change legend title font size
        legend.text = element_text(size=6)) +
  guides(fill = guide_colourbar(title = "Standardized count")) +
  annotate(geom = 'text', label = "b", x = -Inf, y = Inf, hjust = -0.3, vjust = 1, size = 3)

## Before shift
before_shift_plot <- ggplot(data = hex_full_count_df, aes(x = x, y = y)) +
  geom_polygon(color = "black", aes(group = polygon_id, fill = std_counts)) +
  geom_text(aes(x = c_x, y = c_y, label = hexID), size = 2) +
  scale_fill_viridis_c(direction = -1, na.value = "#ffffff", option = "C") +
  coord_equal() +
  theme_void() +
  theme(legend.position="bottom", legend.direction="horizontal", plot.title = element_text(size = 7, hjust = 0.5, vjust = 0.5),
        axis.title.x = element_blank(), axis.title.y = element_blank(),
        axis.text.x = element_blank(), axis.ticks.x = element_blank(),
        axis.text.y = element_blank(), axis.ticks.y = element_blank())

```

```

axis.text.x = element_blank(), axis.ticks.x = element_blank(),
axis.text.y = element_blank(), axis.ticks.y = element_blank(),
panel.grid.major = element_blank(), panel.grid.minor = element_blank(), #change legend key width
legend.title = element_text(size=8), #change legend title font size
legend.text = element_text(size=6)) +
guides(fill = guide_colourbar(title = "Standardized count")) +
annotate(geom = 'text', label = "a", x = -Inf, y = Inf, hjust = -0.3, vjust = 1, size = 3)

## After shift
after_shift_plot <- ggplot(data = shifted_hex_coord_df, aes(x = x, y = y)) +
  geom_polygon(color = "black", aes(group = polygon_id, fill = std_counts)) +
  geom_text(aes(x = c_x, y = c_y, label = hexID), size = 2) +
  scale_fill_viridis_c(direction = -1, na.value = "#ffffff", option = "C") +
  coord_equal() +
  theme_void() +
  theme(legend.position="none", legend.direction="horizontal", plot.title = element_text(size = 7, hjust = 0.5, vjust = 0.5),
        axis.title.x = element_blank(), axis.title.y = element_blank(),
        axis.text.x = element_blank(), axis.ticks.x = element_blank(),
        axis.text.y = element_blank(), axis.ticks.y = element_blank(),
        panel.grid.major = element_blank(), panel.grid.minor = element_blank(), #change legend key width
        legend.title = element_text(size=8), #change legend title font size
        legend.text = element_text(size=6)) +
guides(fill = guide_colourbar(title = "Standardized count")) +
annotate(geom = 'text', label = "b", x = -Inf, y = Inf, hjust = -0.3, vjust = 1, size = 3)

```

Benchmark value to remove the low-density hexagons

```

## As an option first quantile considered as a default
benchmark_to_rm_lwd_hex <- quantile(df_bin_centroids$std_counts)[2] + 0.01

## To identify low density hexagons
df_bin_centroids_low <- df_bin_centroids |>
  dplyr::filter(std_counts <= benchmark_to_rm_lwd_hex)

## To identify low-density hexagons needed to remove by investigating neighbouring mean density
identify_rm_bins <- find_low_density_hexagons(df_bin_centroids_all = df_bin_centroids, num_bins_x = num_bins_x,
                                                df_bin_centroids_low = df_bin_centroids_low)

```

Benchmark value to remove the long edges

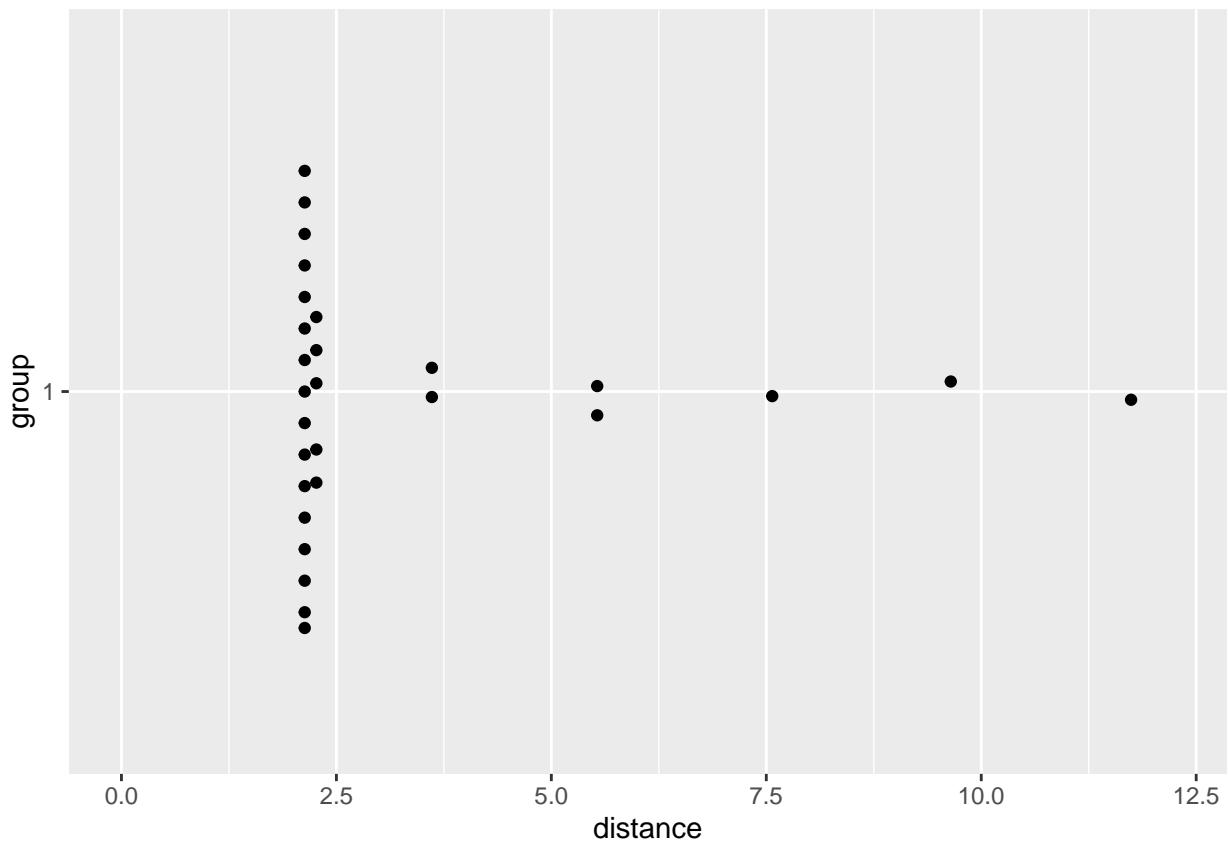
```

## Compute 2D distances
distance <- cal_2d_dist(.data = tr_from_to_df)

## To plot the distribution of distance
plot_dist <- function(distance_df){
  distance_df$group <- "1"
  dist_plot <- ggplot(distance_df, aes(x = group, y = distance)) +
    geom_quasirandom()+
    ylim(0, max(unlist(distance_df$distance))+ 0.5) + coord_flip()
  return(dist_plot)
}

plot_dist(distance)

```



```
benchmark <- find_benchmark_value(.data = distance, distance_col = "distance")
benchmark <- 3
```

Model function

Predict 2D embeddings

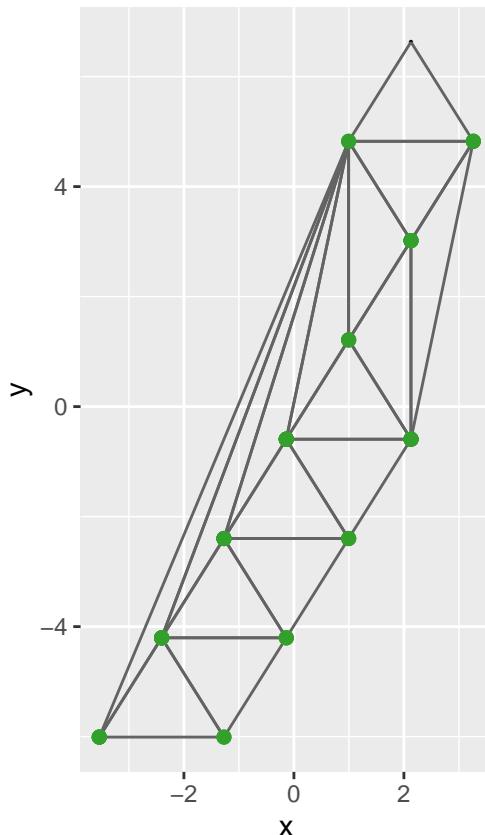
Compute residuals

Visualizations

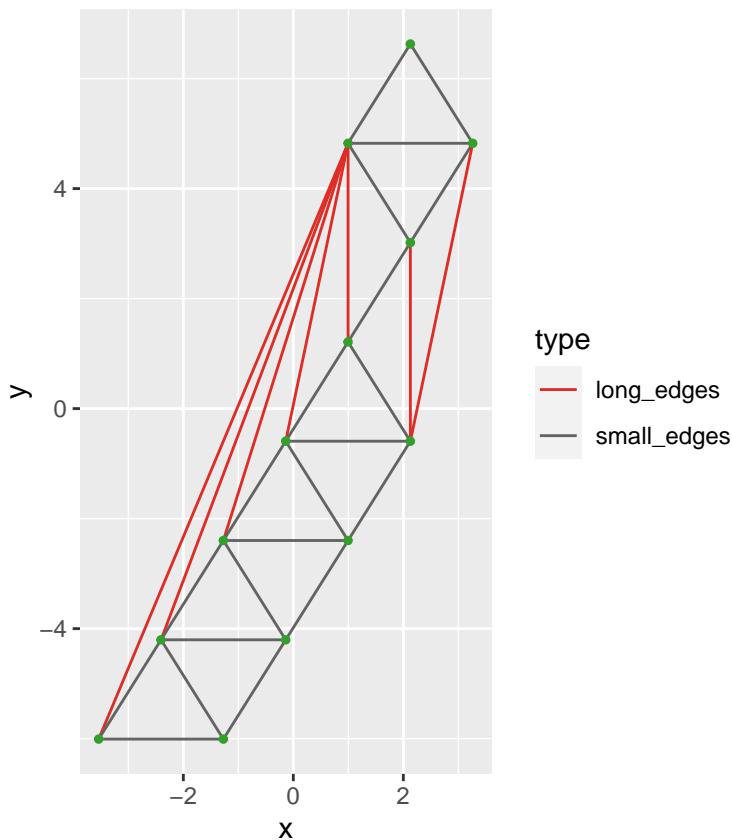
geom_trimesh

```
trimesh <- ggplot(df_bin_centroids, aes(x = x, y = y)) +
  geom_point(size = 0.1) +
  geom_trimesh() +
  coord_equal()

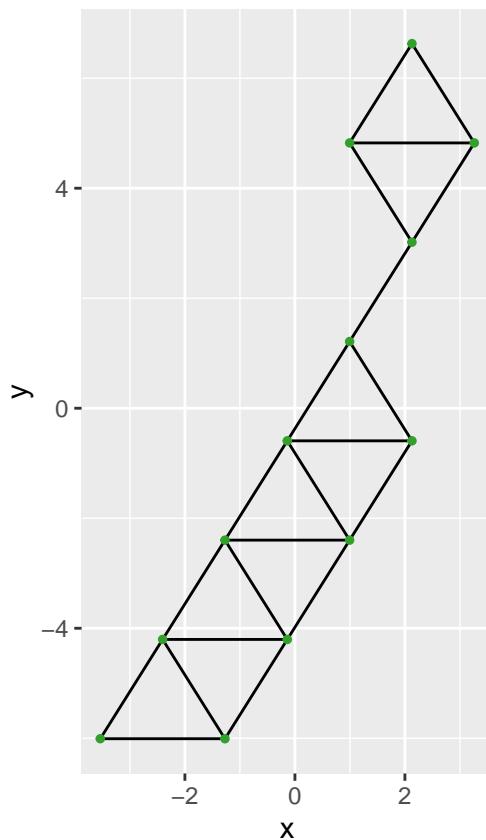
trimesh
```

**coloured_long_edges**

```
trimesh_gr <- colour_long_edges(.data = distance, benchmark_value = benchmark,
                                 triangular_object = tr1_object, distance_col = "distance")  
  
trimesh_gr
```

**remove long edges**

```
trimesh_removed <- remove_long_edges(.data = distance, benchmark_value = benchmark,
                                      triangular_object = tr1_object, distance_col = "distance")
trimesh_removed
```



```
show_langevitour

## To generate a data set with high-D and 2D training data
df_all <- training_data |> dplyr::select(-ID) |>
  dplyr::bind_cols(s_curve_noise_umap_with_id)

## To generate averaged high-D data

df_bin <- avg_highD_data(.data = df_all, column_start_text = "x") ## Need to pass ID column name

tour1 <- show_langevitour(df_all, df_bin, df_bin_centroids, benchmark_value = benchmark,
                           distance = distance, distance_col = "distance")
tour1
```

Tests

All functions have tests written and implemented using the [testthat](#) (Wickham 2011) in R.

3 Application

```

medlea_df <- read_csv("data/medlea_dataset.csv")
names(medlea_df)[2:(NCOL(medlea_df) - 1)] <- paste0("x", 1:(NCOL(medlea_df) - 2))

medlea_df <- medlea_df |> ## Since only contains zeros
  select(-x10)

#medlea_df[,2:(NCOL(medlea_df) - 1)] <- scale(medlea_df[,2:(NCOL(medlea_df) - 1)])

calculate_pca <- function(feature_dataset, num_pcs){
  pcaY_cal <- prcomp(feature_dataset, center = TRUE, scale = TRUE)
  PCAresults <- data.frame(pcaY_cal$x[, 1:num_pcs])
  summary_pca <- summary(pcaY_cal)
  var_explained_df <- data.frame(PC= paste0("PC",1:50),
    var_explained=(pcaY_cal$sdev[1:50])^2/sum((pcaY_cal$sdev[1:50])^2))
  return(list(prcomp_out = pcaY_cal,pca_components = PCAresults, summary = summary_pca, var_explained_pca = var_explained_df))
}

features <- medlea_df[,2:(NCOL(medlea_df) - 1)]
pca_ref_calc <- calculate_pca(features, 8)
pca_ref_calc$summary

#> Importance of components:
#>          PC1     PC2     PC3     PC4     PC5     PC6     PC7
#> Standard deviation 3.1691 3.0609 2.7226 1.87967 1.71219 1.34192 1.27525
#> Proportion of Variance 0.1969 0.1837 0.1453 0.06928 0.05748 0.03531 0.03189
#> Cumulative Proportion 0.1969 0.3806 0.5260 0.59526 0.65274 0.68805 0.71993
#>          PC8     PC9     PC10    PC11    PC12    PC13    PC14
#> Standard deviation 1.16992 1.13465 1.06628 1.03279 0.97899 0.96264 0.9528
#> Proportion of Variance 0.02684 0.02524 0.02229 0.02091 0.01879 0.01817 0.0178
#> Cumulative Proportion 0.74677 0.77202 0.79431 0.81522 0.83402 0.85219 0.8700
#>          PC15    PC16    PC17    PC18    PC19    PC20    PC21
#> Standard deviation 0.9116 0.9090 0.79750 0.76725 0.72414 0.65310 0.61052
#> Proportion of Variance 0.0163 0.0162 0.01247 0.01154 0.01028 0.00836 0.00731
#> Cumulative Proportion 0.8863 0.9025 0.91496 0.92650 0.93678 0.94514 0.95245
#>          PC22    PC23    PC24    PC25    PC26    PC27    PC28
#> Standard deviation 0.6019 0.55399 0.52293 0.46638 0.41959 0.3976 0.34697
#> Proportion of Variance 0.0071 0.00602 0.00536 0.00426 0.00345 0.0031 0.00236
#> Cumulative Proportion 0.9596 0.96557 0.97093 0.97520 0.97865 0.9818 0.98411
#>          PC29    PC30    PC31    PC32    PC33    PC34    PC35
#> Standard deviation 0.33415 0.30618 0.29237 0.28458 0.26033 0.25420 0.22792
#> Proportion of Variance 0.00219 0.00184 0.00168 0.00159 0.00133 0.00127 0.00102
#> Cumulative Proportion 0.98630 0.98814 0.98982 0.99140 0.99273 0.99400 0.99502
#>          PC36    PC37    PC38    PC39    PC40    PC41    PC42
#> Standard deviation 0.21644 0.20437 0.19127 0.1744 0.15586 0.15252 0.12519
#> Proportion of Variance 0.00092 0.00082 0.00072 0.0006 0.00048 0.00046 0.00031
#> Cumulative Proportion 0.99594 0.99676 0.99747 0.9981 0.99855 0.99900 0.99931
#>          PC43    PC44    PC45    PC46    PC47    PC48    PC49
#> Standard deviation 0.10485 0.08598 0.08008 0.06491 0.04841 0.04094 0.03791
#> Proportion of Variance 0.00022 0.00014 0.00013 0.00008 0.00005 0.00003 0.00003
#> Cumulative Proportion 0.99952 0.99967 0.99980 0.99988 0.99992 0.99996 0.99999
#>          PC50    PC51
#> Standard deviation 0.02347 0.01421
#> Proportion of Variance 0.00001 0.00000
#> Cumulative Proportion 1.00000 1.00000

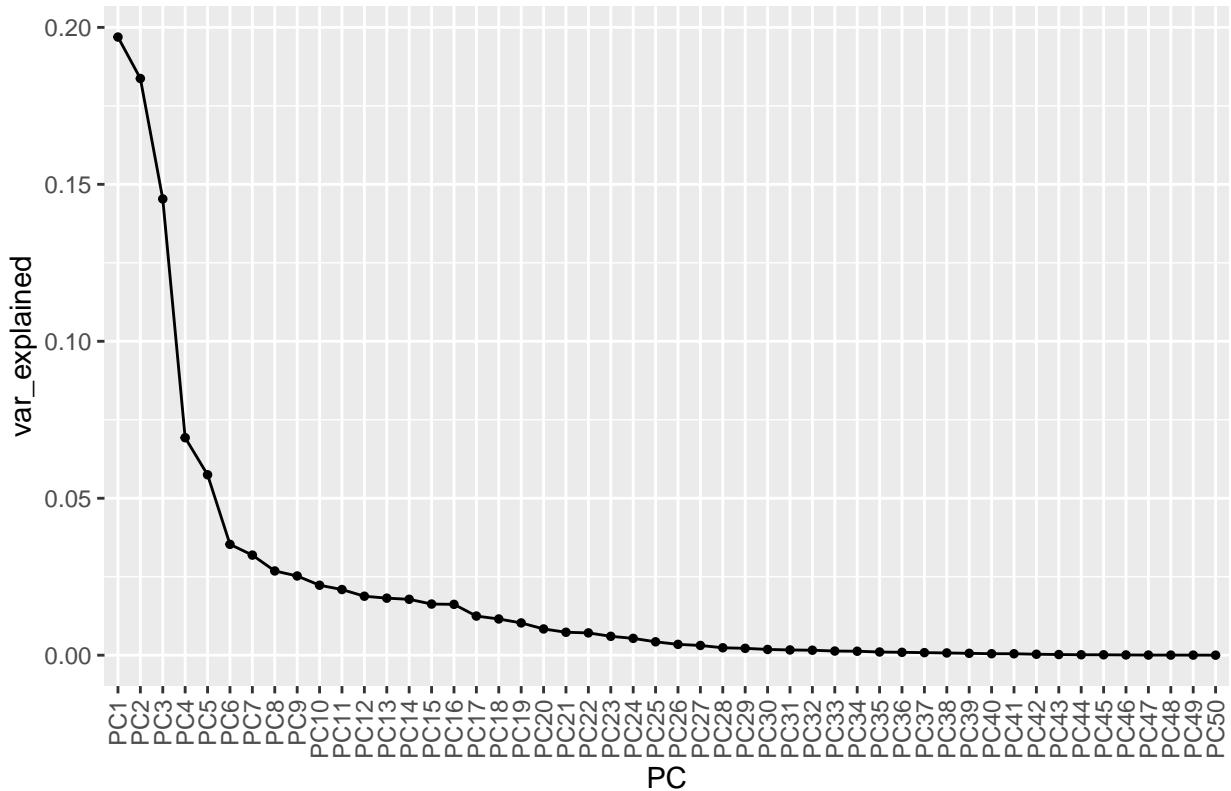
var_explained_df <- pca_ref_calc$var_explained_pca
data_pca <- pca_ref_calc$pca_components |>
  mutate(ID = 1:NROW(pca_ref_calc$pca_components),
    shape_label = medlea_df$Shape_label)

var_explained_df |>
  ggplot(aes(x = PC,y = var_explained, group = 1))+
  geom_point(size=1)

```

```
geom_line()+
  labs(title="Scree plot: PCA on scaled data") +
  scale_x_discrete(limits = paste0(rep("PC", 50), 1:50)) +
  theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1))
```

Scree plot: PCA on scaled data



```
data_split <- initial_split(data_pca)
training_data <- training(data_split) |>
  arrange(ID)
test_data <- testing(data_split) |>
  arrange(ID)

UMAP_fit <- umap(training_data |> dplyr::select(-c(ID, shape_label)), n_neighbors = 37, n_components = 2)

UMAP_data <- UMAP_fit$layout |>
  as.data.frame()
names(UMAP_data)[1:(ncol(UMAP_data))] <- paste0(rep("UMAP", (ncol(UMAP_data))), 1:(ncol(UMAP_data)))

UMAP_data <- UMAP_data |>
  mutate(ID = training_data$id)

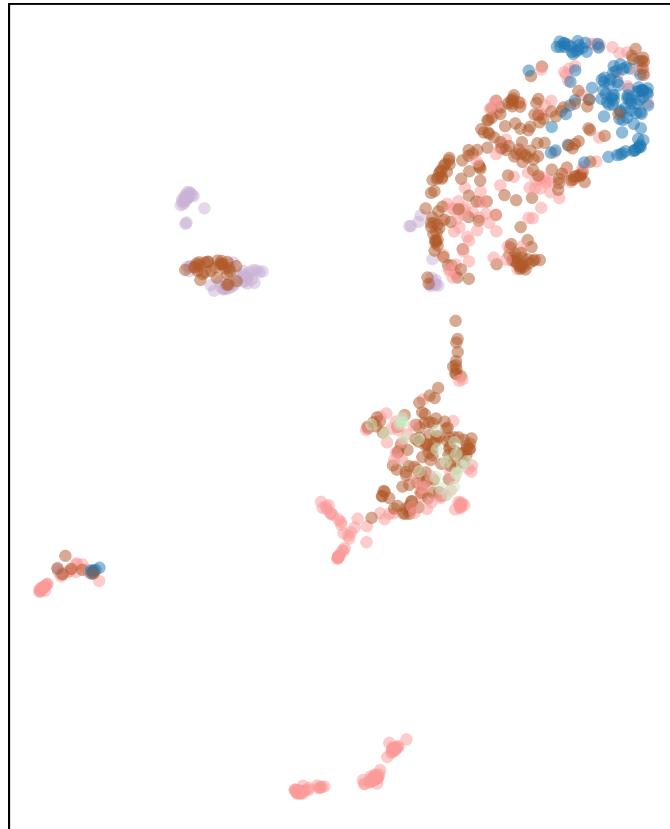
UMAP_data_with_label <- UMAP_data |>
  mutate(shape_label = training_data$shape_label)

UMAP_data_with_label |>
  ggplot(aes(x = UMAP1,
             y = UMAP2, color = shape_label)) +
  geom_point(alpha=0.5) +
  coord_equal() +
  theme(plot.title = element_text(hjust = 0.5, size = 18, face = "bold")) + #ggtitle("(a)") +
  theme_linedraw() +
  theme(legend.position = "none", plot.title = element_text(size = 7, hjust = 0.5, vjust = -0.5),
        axis.title.x = element_blank(), axis.title.y = element_blank(),
        axis.text.x = element_blank(), axis.ticks.x = element_blank(),
        axis.text.y = element_blank(), axis.ticks.y = element_blank(),
```

```

panel.grid.major = element_blank(), panel.grid.minor = element_blank(), #change legend key width
legend.title = element_text(size=5), #change legend title font size
legend.text = element_text(size=4),
  legend.key.height = unit(0.25, 'cm'),
  legend.key.width = unit(0.25, 'cm')) +
scale_color_manual(values=c("#b15928", "#1f78b4", "#cab2d6", "#ccebc5", "#fb9a99", "#e31a1c", "#6a3d9a", "#fff7

```



```

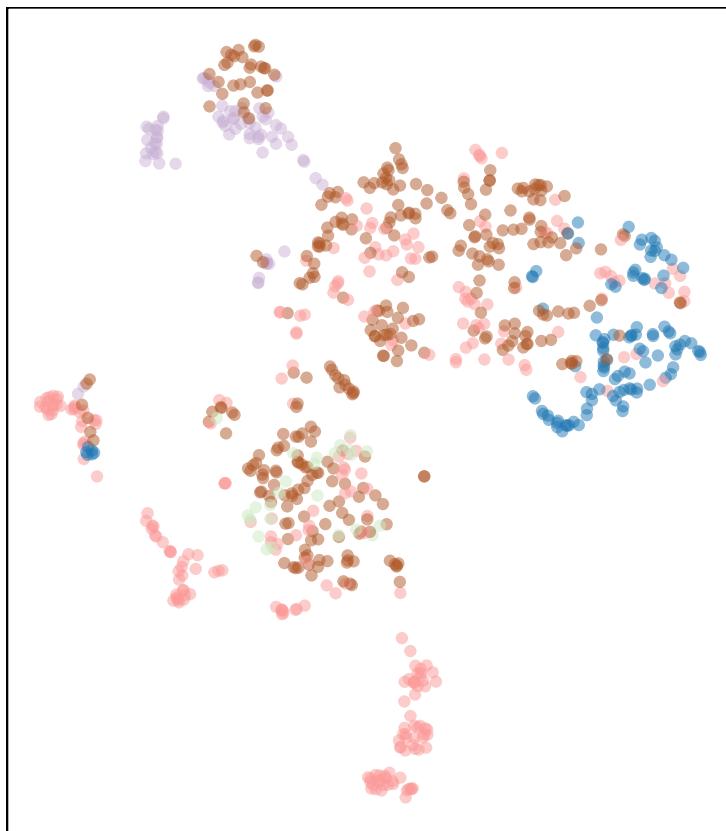
tsNE_data <- Fit_tSNE(training_data |> dplyr::select(-c(ID, shape_label)), opt_perplexity = calculate_effective

tsNE_data <- tsNE_data |>
  select(-ID) |>
  mutate(ID = training_data$ID)

tsNE_data_with_label <- tsNE_data |>
  mutate(shape_label = training_data$shape_label)

tsNE_data_with_label |>
  ggplot(aes(x = tSNE1,
             y = tSNE2, color = shape_label))+
    geom_point(alpha=0.5) +
    coord_equal() +
    theme(plot.title = element_text(hjust = 0.5, size = 18, face = "bold")) + #ggtitle("(a)") +
    theme_linedraw() +
    theme(legend.position = "none", plot.title = element_text(size = 7, hjust = 0.5, vjust = -0.5),
          axis.title.x = element_blank(), axis.title.y = element_blank(),
          axis.text.x = element_blank(), axis.ticks.x = element_blank(),
          axis.text.y = element_blank(), axis.ticks.y = element_blank(),
          panel.grid.major = element_blank(), panel.grid.minor = element_blank(), #change legend key width
          legend.title = element_text(size=5), #change legend title font size
          legend.text = element_text(size=4),
          legend.key.height = unit(0.25, 'cm'),
          legend.key.width = unit(0.25, 'cm')) +
        scale_color_manual(values=c("#b15928", "#1f78b4", "#cab2d6", "#ccebc5", "#fb9a99", "#e31a1c", "#6a3d9a", "#fff7

```



```

PHATE_data <- Fit_PHATE(training_data |> dplyr::select(-c(ID, shape_label)), knn = 5, with_seed = 20240110)

#> Calculating PHATE...
#>   Running PHATE on 824 observations and 8 variables.
#>   Calculating graph and diffusion operator...
#>     Calculating KNN search...
#>     Calculating affinities...
#>   Calculated graph and diffusion operator in 0.01 seconds.
#>   Calculating optimal t...
#>     Automatically selected t = 22
#>   Calculated optimal t in 0.26 seconds.
#>   Calculating diffusion potential...
#>   Calculated diffusion potential in 0.36 seconds.
#>   Calculating metric MDS...
#>   Calculated metric MDS in 8.01 seconds.
#>   Calculated PHATE in 8.64 seconds.

PHATE_data <- PHATE_data |>
  select(PHATE1, PHATE2)
PHATE_data <- PHATE_data |>
  mutate(ID = training_data$ID)

PHATE_data_with_label <- PHATE_data |>
  mutate(shape_label = training_data$shape_label)

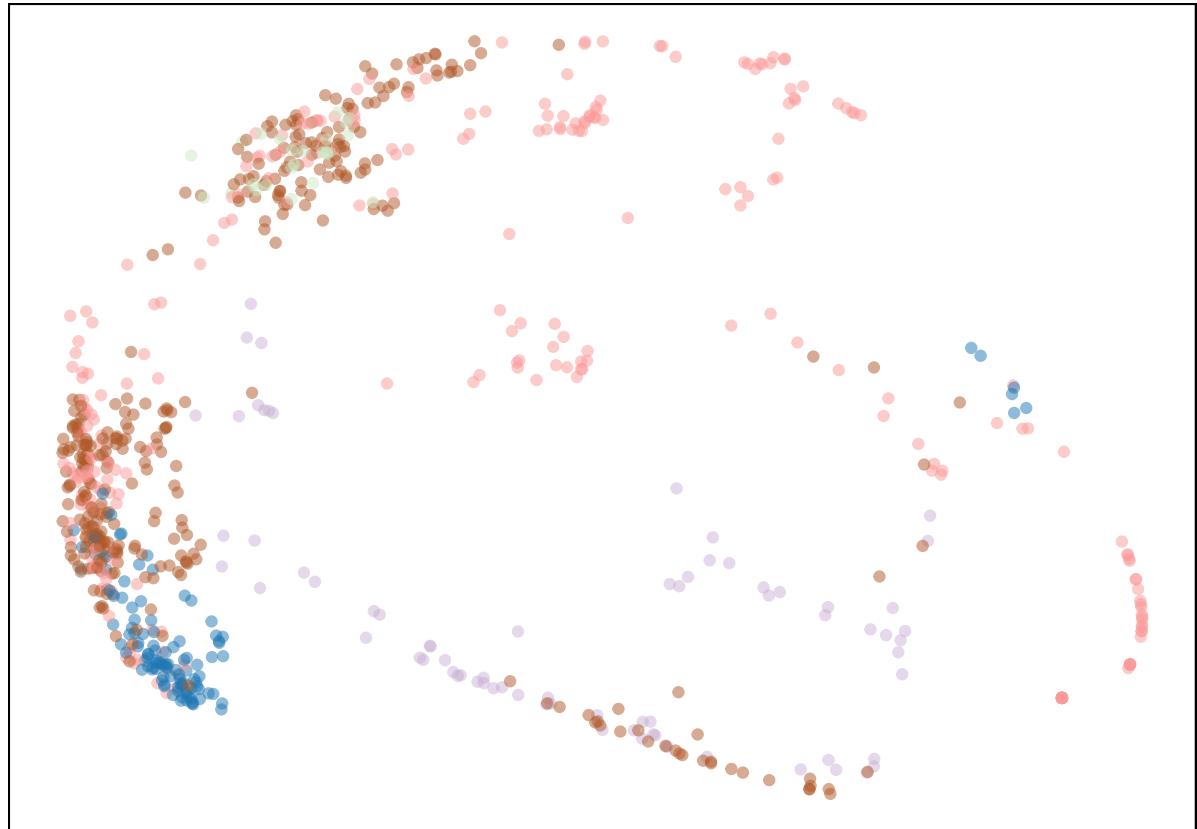
PHATE_data_with_label |>
  ggplot(aes(x = PHATE1,
             y = PHATE2, color = shape_label)) +
  geom_point(alpha=0.5) +
  coord_equal() +
  theme(plot.title = element_text(hjust = 0.5, size = 18, face = "bold")) + #ggtitle("(a)") +
  theme_linedraw() +
  theme(legend.position = "none", plot.title = element_text(size = 7, hjust = 0.5, vjust = -0.5),
        axis.title.x = element_blank(), axis.title.y = element_blank(),
        axis.text.x = element_blank(), axis.ticks.x = element_blank(),
        axis.text.y = element_blank())

```

```

axis.text.y = element_blank(), axis.ticks.y = element_blank(),
panel.grid.major = element_blank(), panel.grid.minor = element_blank(), #change legend key width
legend.title = element_text(size=5), #change legend title font size
legend.text = element_text(size=4),
legend.key.height = unit(0.25, 'cm'),
legend.key.width = unit(0.25, 'cm')) +
scale_color_manual(values=c("#b15928", "#1f78b4", "#cab2d6", "#ccebc5", "#fb9a99", "#e31a1c", "#6a3d9a", "#ff7f0e"))

```



```

tem_dir <- tempdir()

Fit_TriMAP_data(training_data |> dplyr::select(-c(ID, shape_label)), tem_dir)

path <- file.path(tem_dir, "df_2_without_class.csv")
path2 <- file.path(tem_dir, "dataset_3_TriMAP_values.csv")

Fit_TriMAP(as.integer(2), as.integer(5), as.integer(4), as.integer(3), path, path2)

TriMAP_data <- read_csv(path2)
TriMAP_data <- TriMAP_data |>
  mutate(ID = training_data$ID)

TriMAP_data_with_label <- TriMAP_data |>
  mutate(shape_label = training_data$shape_label)

TriMAP_data_with_label |>
  ggplot(aes(x = TriMAP1,
             y = TriMAP2, color = shape_label)) +
  geom_point(alpha=0.5) +
  coord_equal() +
  theme(plot.title = element_text(hjust = 0.5, size = 18, face = "bold")) + #ggtitle("(a)") +
  theme_linedraw() +
  theme(legend.position = "none", plot.title = element_text(size = 7, hjust = 0.5, vjust = -0.5),
        axis.title.x = element_blank(), axis.title.y = element_blank(),
        axis.text.x = element_blank(), axis.ticks.x = element_blank(),
        axis.text.y = element_blank(), axis.ticks.y = element_blank(),

```

```

panel.grid.major = element_blank(), panel.grid.minor = element_blank(), #change legend key width
legend.title = element_text(size=5), #change legend title font size
legend.text = element_text(size=4),
  legend.key.height = unit(0.25, 'cm'),
  legend.key.width = unit(0.25, 'cm')) +
scale_color_manual(values=c("#b15928", "#1f78b4", "#cab2d6", "#ccebc5", "#fb9a99", "#e31a1c", "#6a3d9a", "#ff7f0e"))

```



```

tem_dir <- tempdir()

Fit_PacMAP_data(training_data |> dplyr::select(-c(ID, shape_label)), tem_dir)

path <- file.path(tem_dir, "df_2_without_class.csv")
path2 <- file.path(tem_dir, "dataset_3_PaCMAP_values.csv")

Fit_PaCMAP(as.integer(2), as.integer(10), "random", 0.9, as.integer(2), path, path2)

PaCMAP_data <- read_csv(path2)
PaCMAP_data <- PaCMAP_data |>
  mutate(ID = training_data$ID)

PaCMAP_data_with_label <- PaCMAP_data |>
  mutate(shape_label = training_data$shape_label)

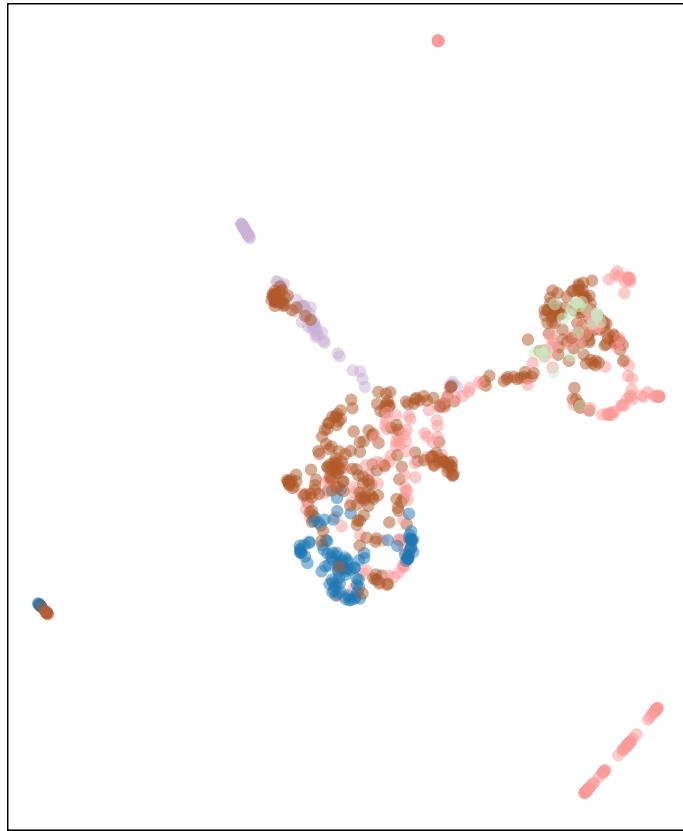
PaCMAP_data_with_label |>
  ggplot(aes(x = PaCMAP1,
             y = PaCMAP2, color = shape_label))+
  geom_point(alpha=0.5) +
  coord_equal() +
  theme(plot.title = element_text(hjust = 0.5, size = 18, face = "bold")) + #ggtitle("(a)") +
  theme_linedraw() +
  theme(legend.position = "none", plot.title = element_text(size = 7, hjust = 0.5, vjust = -0.5),
        axis.title.x = element_blank(), axis.title.y = element_blank(),
        axis.text.x = element_blank(), axis.ticks.x = element_blank(),
        axis.text.y = element_blank(), axis.ticks.y = element_blank(),
        panel.grid.major = element_blank(), panel.grid.minor = element_blank(), #change legend key width

```

```

legend.title = element_text(size=5), #change legend title font size
legend.text = element_text(size=4),
  legend.key.height = unit(0.25, 'cm'),
  legend.key.width = unit(0.25, 'cm')) +
scale_color_manual(values=c("#b15928", "#1f78b4", "#cab2d6", "#ccebc5", "#fb9a99", "#e31a1c", "#6a3d9a", "#ff7f0e"))

```



```

num_bins_x <- calculate_effective_x_bins(.data = tSNE_data, x = "tSNE1", hex_size = NA)

num_bins_y <- calculate_effective_y_bins(.data = tSNE_data, y = "tSNE2", hex_size = NA)
num_bins_y

#> [1] 38

all_centroids_df <- generate_full_grid_centroids(nldr_df = tSNE_data,
                                                    x = "tSNE1", y = "tSNE2",
                                                    num_bins_x = num_bins_x,
                                                    num_bins_y = num_bins_y,
                                                    buffer_size = NA, hex_size = NA)

hex_grid <- gen_hex_coordinates(all_centroids_df)

full_grid_with_hexbin_id <- map_hexbin_id(all_centroids_df)

full_grid_with_polygon_id <- map_polygon_id(full_grid_with_hexbin_id, hex_grid)

tSNE_data_with_id <- assign_data(tSNE_data, full_grid_with_hexbin_id)

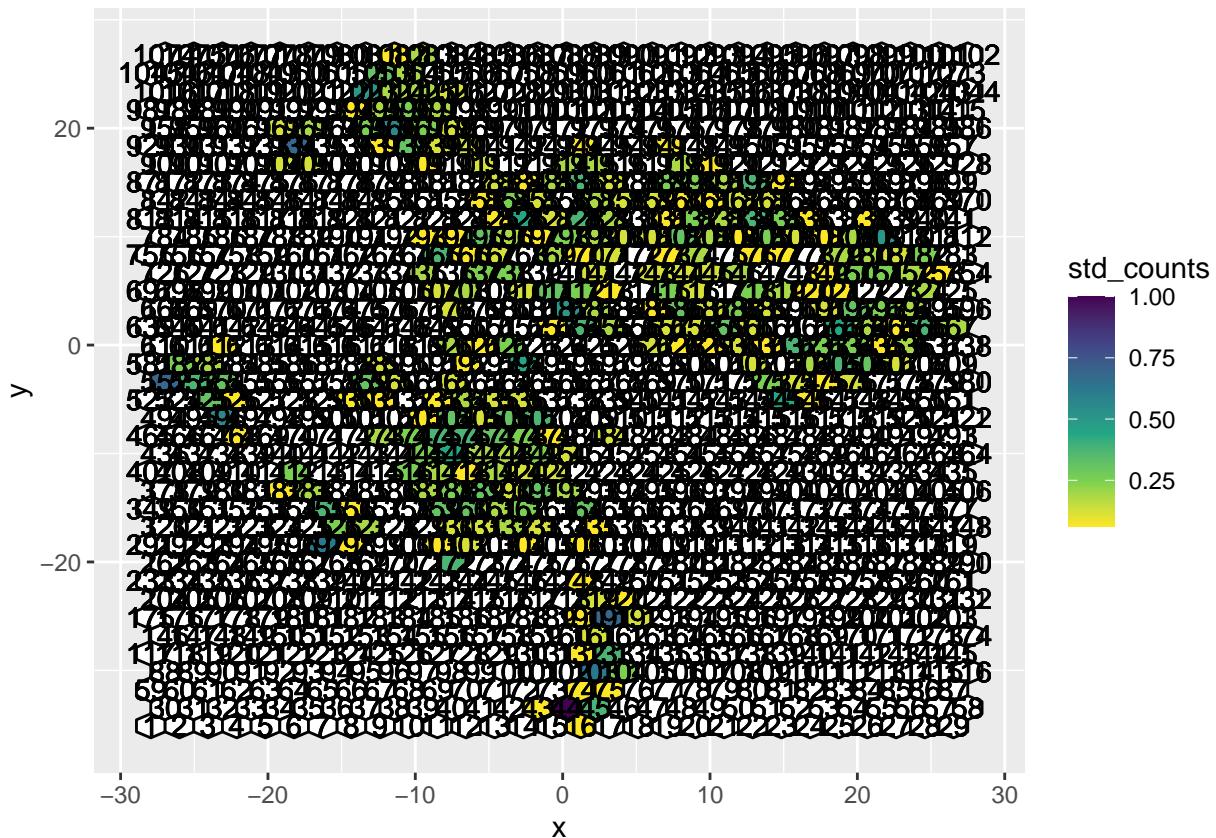
df_with_std_counts <- compute_std_counts(nldr_df = tSNE_data_with_id)

hex_full_count_df <- generate_full_grid_info(full_grid_with_polygon_id, df_with_std_counts, hex_grid)

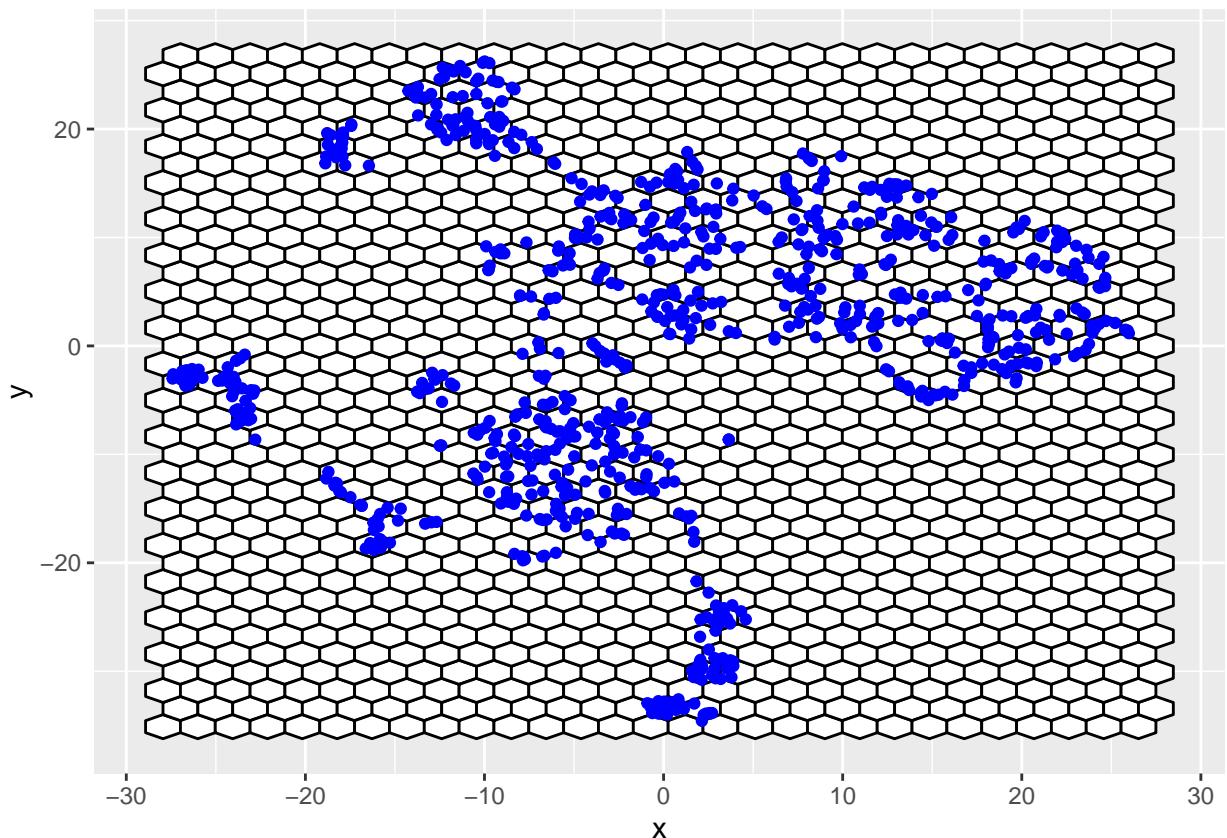
ggplot(data = hex_full_count_df, aes(x = x, y = y)) +
  geom_polygon(color = "black", aes(group = polygon_id, fill = std_counts)) +

```

```
geom_text(aes(x = c_x, y = c_y, label = hexID)) +  
scale_fill_viridis_c(direction = -1, na.value = "#ffffff")
```



```
ggplot(data = hex_grid, aes(x = x, y = y)) + geom_polygon(fill = "white", color = "black", aes(group = id)) +
  geom_point(data = tSNE_data, aes(x = tSNE1, y = tSNE2), color = "blue")
```



```

df_bin_centroids <- hex_full_count_df[complete.cases(hex_full_count_df[["std_counts"]]), ] |>
  dplyr::select("c_x", "c_y", "hexID", "std_counts") |>
  dplyr::distinct() |>
  dplyr::rename(c("x" = "c_x", "y" = "c_y"))

df_bin_centroids

#>           x         y   hexID std_counts
#> 1 -26.9783746 -3.336399929    552  0.6875
#> 2 -25.0331153 -3.336399929    553  0.3750
#> 3 -26.0057449 -1.663988875    582  0.2500
#> 4 -23.0878560 -6.681222036    496  0.5625
#> 5 -24.0604856 -5.008810982    525  0.3125
#> 6 -23.0878560 -3.336399929    554  0.3125
#> 7 -24.0604856 -1.663988875    583  0.1875
#> 8 -23.0878560  0.008422179    612  0.0625
#> 9 -22.1152263 -8.353633090    468  0.0625
#> 10 -22.1152263 -5.008810982   526  0.0625
#> 11 -19.1973374 -13.370866252   382  0.0625
#> 12 -19.1973374  16.732532718   904  0.1250
#> 13 -19.1973374  20.077354826   962  0.1250
#> 14 -17.2520781 -13.370866252   383  0.1875
#> 15 -18.2247077 -11.698455198   412  0.2500
#> 16 -17.2520781  16.732532718   905  0.1250
#> 17 -18.2247077  18.404943772   934  0.6875
#> 18 -17.2520781  20.077354826   963  0.1875
#> 19 -16.2794484 -18.388099413   297  0.6250
#> 20 -15.3068188 -16.715688359   326  0.2500
#> 21 -16.2794484 -15.043277306   355  0.3750
#> 22 -14.3341891 -18.388099413   298  0.0625
#> 23 -13.3615595 -16.715688359   327  0.1875
#> 24 -14.3341891 -15.043277306   356  0.0625
#> 25 -14.3341891 -5.008810982   530  0.0625
#> 26 -13.3615595 -3.336399929   559  0.3125
#> 27 -13.3615595  20.077354826   965  0.2500
#> 28 -14.3341891  21.749765879   994  0.0625
#> 29 -13.3615595  23.422176933  1023  0.5000
#> 30 -12.3889298 -8.353633090   473  0.1875
#> 31 -12.3889298 -5.008810982   531  0.0625
#> 32 -11.4163002 -3.336399929   560  0.1250
#> 33 -12.3889298 -1.663988875   589  0.1250
#> 34 -12.3889298  18.404943772   937  0.0625
#> 35 -11.4163002  20.077354826   966  0.5625
#> 36 -12.3889298  21.749765879   995  0.1875
#> 37 -11.4163002  23.422176933  1024  0.1875
#> 38 -12.3889298  25.094587987  1053  0.3125
#> 39 -11.4163002  26.766999041  1082  0.0625
#> 40 -9.4710409 -13.370866252   387  0.1875
#> 41 -10.4436705 -11.698455198   416  0.2500
#> 42 -9.4710409 -10.026044144   445  0.2500
#> 43 -10.4436705 -8.353633090   474  0.2500
#> 44 -9.4710409 -6.681222036   503  0.0625
#> 45 -9.4710409  6.698066395   735  0.1875
#> 46 -9.4710409  10.042888502   793  0.0625
#> 47 -9.4710409  16.732532718   909  0.0625
#> 48 -10.4436705  18.404943772   938  0.3125
#> 49 -9.4710409  20.077354826   967  0.2500
#> 50 -10.4436705  21.749765879   996  0.2500
#> 51 -9.4710409  23.422176933  1025  0.1875
#> 52 -10.4436705  25.094587987  1054  0.1875
#> 53 -9.4710409  26.766999041  1083  0.1875
#> 54 -7.5257816 -20.060510467   272  0.3750
#> 55 -8.4984112 -18.388099413   301  0.0625
#> 56 -7.5257816 -16.715688359   330  0.1250

```

```
#> 57 -8.4984112 -15.043277306 359 0.3125
#> 58 -7.5257816 -13.370866252 388 0.2500
#> 59 -8.4984112 -11.698455198 417 0.1875
#> 60 -7.5257816 -10.026044144 446 0.4375
#> 61 -8.4984112 -8.353633090 475 0.3750
#> 62 -7.5257816 -6.681222036 504 0.3125
#> 63 -8.4984112 -5.008810982 533 0.0625
#> 64 -7.5257816 -3.336399929 562 0.1250
#> 65 -7.5257816 0.008422179 620 0.1875
#> 66 -7.5257816 3.353244287 678 0.1250
#> 67 -8.4984112 5.025655341 707 0.1250
#> 68 -8.4984112 8.370477449 765 0.3750
#> 69 -7.5257816 10.042888502 794 0.0625
#> 70 -8.4984112 18.404943772 939 0.1250
#> 71 -7.5257816 20.077354826 968 0.1250
#> 72 -8.4984112 21.749765879 997 0.1875
#> 73 -7.5257816 23.422176933 1026 0.1250
#> 74 -6.5531519 -18.388099413 302 0.1250
#> 75 -5.5805223 -16.715688359 331 0.1250
#> 76 -6.5531519 -15.043277306 360 0.3125
#> 77 -5.5805223 -13.370866252 389 0.3125
#> 78 -6.5531519 -11.698455198 418 0.0625
#> 79 -5.5805223 -10.026044144 447 0.1875
#> 80 -6.5531519 -8.353633090 476 0.3750
#> 81 -5.5805223 -6.681222036 505 0.1250
#> 82 -6.5531519 -5.008810982 534 0.1875
#> 83 -6.5531519 -1.663988875 592 0.0625
#> 84 -5.5805223 0.008422179 621 0.0625
#> 85 -6.5531519 5.025655341 708 0.1875
#> 86 -5.5805223 6.698066395 737 0.2500
#> 87 -6.5531519 8.370477449 766 0.0625
#> 88 -5.5805223 10.042888502 795 0.3125
#> 89 -5.5805223 13.387710610 853 0.0625
#> 90 -5.5805223 16.732532718 911 0.1250
#> 91 -6.5531519 18.404943772 940 0.1250
#> 92 -3.6352630 -16.715688359 332 0.1875
#> 93 -4.6078926 -15.043277306 361 0.1875
#> 94 -3.6352630 -13.370866252 390 0.1250
#> 95 -4.6078926 -11.698455198 419 0.1875
#> 96 -3.6352630 -10.026044144 448 0.1875
#> 97 -4.6078926 -8.353633090 477 0.3125
#> 98 -3.6352630 -6.681222036 506 0.3125
#> 99 -4.6078926 -5.008810982 535 0.1250
#> 100 -3.6352630 0.008422179 622 0.2500
#> 101 -3.6352630 6.698066395 738 0.2500
#> 102 -4.6078926 8.370477449 767 0.1250
#> 103 -3.6352630 10.042888502 796 0.1875
#> 104 -4.6078926 11.715299556 825 0.0625
#> 105 -3.6352630 13.387710610 854 0.2500
#> 106 -4.6078926 15.060121664 883 0.1250
#> 107 -1.6900037 -33.439798898 43 0.0625
#> 108 -2.6626333 -18.388099413 304 0.0625
#> 109 -1.6900037 -16.715688359 333 0.1250
#> 110 -2.6626333 -15.043277306 362 0.2500
#> 111 -1.6900037 -13.370866252 391 0.3125
#> 112 -2.6626333 -11.698455198 420 0.1250
#> 113 -1.6900037 -10.026044144 449 0.2500
#> 114 -2.6626333 -8.353633090 478 0.2500
#> 115 -1.6900037 -6.681222036 507 0.3750
#> 116 -2.6626333 -5.008810982 536 0.1250
#> 117 -2.6626333 -1.663988875 594 0.4375
#> 118 -2.6626333 5.025655341 710 0.1250
#> 119 -1.6900037 10.042888502 797 0.0625
#> 120 -2.6626333 11.715299556 826 0.4375
```

```
#> 121 -1.6900037 13.387710610 855 0.1250
#> 122 -2.6626333 15.060121664 884 0.0625
#> 123 0.2552556 -33.439798898 44 1.0000
#> 124 0.2552556 -13.370866252 392 0.1875
#> 125 -0.7173740 -11.698455198 421 0.1250
#> 126 0.2552556 -10.026044144 450 0.1250
#> 127 -0.7173740 -8.353633090 479 0.0625
#> 128 -0.7173740 1.680833233 653 0.0625
#> 129 0.2552556 3.353244287 682 0.5000
#> 130 -0.7173740 5.025655341 711 0.3125
#> 131 -0.7173740 8.370477449 769 0.1875
#> 132 0.2552556 10.042888502 798 0.2500
#> 133 -0.7173740 11.715299556 827 0.1250
#> 134 -0.7173740 15.060121664 885 0.1875
#> 135 0.2552556 16.732532718 914 0.1875
#> 136 1.2278853 -35.112209952 16 0.0625
#> 137 2.2005149 -33.439798898 45 0.3750
#> 138 1.2278853 -31.767387844 74 0.0625
#> 139 2.2005149 -30.094976790 103 0.6250
#> 140 1.2278853 -28.422565737 132 0.0625
#> 141 2.2005149 -26.750154683 161 0.1250
#> 142 1.2278853 -25.077743629 190 0.0625
#> 143 2.2005149 -23.405332575 219 0.1250
#> 144 1.2278853 -21.732921521 248 0.0625
#> 145 1.2278853 -18.388099413 306 0.0625
#> 146 2.2005149 -16.715688359 335 0.0625
#> 147 1.2278853 -15.043277306 364 0.3125
#> 148 1.2278853 1.680833233 654 0.3750
#> 149 2.2005149 3.353244287 683 0.1875
#> 150 1.2278853 5.025655341 712 0.2500
#> 151 2.2005149 6.698066395 741 0.1250
#> 152 1.2278853 8.370477449 770 0.0625
#> 153 2.2005149 10.042888502 799 0.1875
#> 154 1.2278853 11.715299556 828 0.3750
#> 155 2.2005149 13.387710610 857 0.1875
#> 156 1.2278853 15.060121664 886 0.3750
#> 157 2.2005149 16.732532718 915 0.1875
#> 158 1.2278853 18.404943772 944 0.0625
#> 159 3.1731446 -31.767387844 75 0.0625
#> 160 4.1457742 -30.094976790 104 0.2500
#> 161 3.1731446 -28.422565737 133 0.3750
#> 162 3.1731446 -25.077743629 191 0.6875
#> 163 4.1457742 -23.405332575 220 0.0625
#> 164 3.1731446 -8.353633090 481 0.1250
#> 165 3.1731446 1.680833233 655 0.1875
#> 166 3.1731446 5.025655341 713 0.0625
#> 167 3.1731446 8.370477449 771 0.1250
#> 168 4.1457742 10.042888502 800 0.1250
#> 169 3.1731446 11.715299556 829 0.1875
#> 170 4.1457742 13.387710610 858 0.1250
#> 171 3.1731446 15.060121664 887 0.1250
#> 172 5.1184039 -25.077743629 192 0.1250
#> 173 6.0910335 0.008422179 627 0.1250
#> 174 6.0910335 3.353244287 685 0.0625
#> 175 6.0910335 6.698066395 743 0.1250
#> 176 6.0910335 10.042888502 801 0.1250
#> 177 6.0910335 13.387710610 859 0.1250
#> 178 8.0362928 0.008422179 628 0.0625
#> 179 7.0636632 1.680833233 657 0.1250
#> 180 8.0362928 3.353244287 686 0.3125
#> 181 7.0636632 5.025655341 715 0.2500
#> 182 8.0362928 6.698066395 744 0.1250
#> 183 7.0636632 8.370477449 773 0.0625
#> 184 8.0362928 10.042888502 802 0.3125
```

```
#> 185 7.0636632 11.715299556 831 0.0625
#> 186 8.0362928 13.387710610 860 0.0625
#> 187 7.0636632 15.060121664 889 0.2500
#> 188 8.0362928 16.732532718 918 0.1875
#> 189 7.0636632 18.404943772 947 0.0625
#> 190 9.9815521 0.008422179 629 0.0625
#> 191 9.0089225 1.680833233 658 0.1250
#> 192 9.9815521 3.353244287 687 0.1250
#> 193 9.0089225 5.025655341 716 0.1250
#> 194 9.9815521 6.698066395 745 0.1250
#> 195 9.0089225 8.370477449 774 0.1875
#> 196 9.9815521 10.042888502 803 0.1875
#> 197 9.0089225 11.715299556 832 0.2500
#> 198 9.9815521 13.387710610 861 0.0625
#> 199 9.0089225 15.060121664 890 0.2500
#> 200 9.9815521 16.732532718 919 0.0625
#> 201 11.9268114 0.008422179 630 0.1250
#> 202 10.9541818 1.680833233 659 0.3125
#> 203 11.9268114 3.353244287 688 0.2500
#> 204 11.9268114 6.698066395 746 0.1875
#> 205 11.9268114 10.042888502 804 0.0625
#> 206 10.9541818 11.715299556 833 0.1875
#> 207 11.9268114 13.387710610 862 0.1250
#> 208 10.9541818 15.060121664 891 0.1875
#> 209 13.8720707 -3.336399929 573 0.2500
#> 210 12.8994411 -1.663988875 602 0.1250
#> 211 13.8720707 0.008422179 631 0.0625
#> 212 12.8994411 1.680833233 660 0.1250
#> 213 13.8720707 3.353244287 689 0.1250
#> 214 12.8994411 5.025655341 718 0.2500
#> 215 12.8994411 8.370477449 776 0.0625
#> 216 13.8720707 10.042888502 805 0.2500
#> 217 12.8994411 11.715299556 834 0.3750
#> 218 13.8720707 13.387710610 863 0.1250
#> 219 12.8994411 15.060121664 892 0.3750
#> 220 14.8447004 -5.008810982 545 0.4375
#> 221 15.8173300 -3.336399929 574 0.1250
#> 222 15.8173300 0.008422179 632 0.3750
#> 223 14.8447004 5.025655341 719 0.1875
#> 224 14.8447004 8.370477449 777 0.0625
#> 225 15.8173300 10.042888502 806 0.1250
#> 226 14.8447004 11.715299556 835 0.2500
#> 227 14.8447004 15.060121664 893 0.0625
#> 228 16.7899597 -5.008810982 546 0.0625
#> 229 17.7625893 -3.336399929 575 0.0625
#> 230 16.7899597 -1.663988875 604 0.1875
#> 231 17.7625893 0.008422179 633 0.2500
#> 232 17.7625893 3.353244287 691 0.1250
#> 233 16.7899597 5.025655341 720 0.0625
#> 234 17.7625893 6.698066395 749 0.0625
#> 235 17.7625893 10.042888502 807 0.0625
#> 236 16.7899597 11.715299556 836 0.0625
#> 237 19.7078486 -3.336399929 576 0.1250
#> 238 18.7352190 -1.663988875 605 0.3750
#> 239 19.7078486 0.008422179 634 0.2500
#> 240 18.7352190 1.680833233 663 0.4375
#> 241 19.7078486 3.353244287 692 0.3125
#> 242 18.7352190 5.025655341 721 0.0625
#> 243 19.7078486 6.698066395 750 0.2500
#> 244 18.7352190 8.370477449 779 0.1875
#> 245 19.7078486 10.042888502 808 0.1875
#> 246 20.6804783 -1.663988875 606 0.3125
#> 247 21.6531079 0.008422179 635 0.0625
#> 248 20.6804783 1.680833233 664 0.1875
```

```

#> 249 21.6531079 3.353244287 693 0.2500
#> 250 21.6531079 6.698066395 751 0.2500
#> 251 20.6804783 8.370477449 780 0.1875
#> 252 21.6531079 10.042888502 809 0.5000
#> 253 20.6804783 11.715299556 838 0.0625
#> 254 22.6257376 -1.663988875 607 0.1250
#> 255 23.5983672 0.008422179 636 0.1875
#> 256 22.6257376 1.680833233 665 0.0625
#> 257 23.5983672 3.353244287 694 0.1875
#> 258 23.5983672 6.698066395 752 0.1875
#> 259 22.6257376 8.370477449 781 0.1250
#> 260 24.5709969 1.680833233 666 0.3750
#> 261 24.5709969 5.025655341 724 0.1875
#> 262 25.5436265 6.698066395 753 0.0625
#> 263 24.5709969 8.370477449 782 0.2500
#> 264 26.5162562 1.680833233 667 0.1875

tr1_object <- triangulate_bin_centroids(df_bin_centroids, x, y)
tr_from_to_df <- generate_edge_info(triangular_object = tr1_object)

## To generate a data set with high-D and 2D training data
df_all <- training_data |> dplyr::select(-c(ID, shape_label)) |>
  dplyr::bind_cols(tSNE_data_with_id)

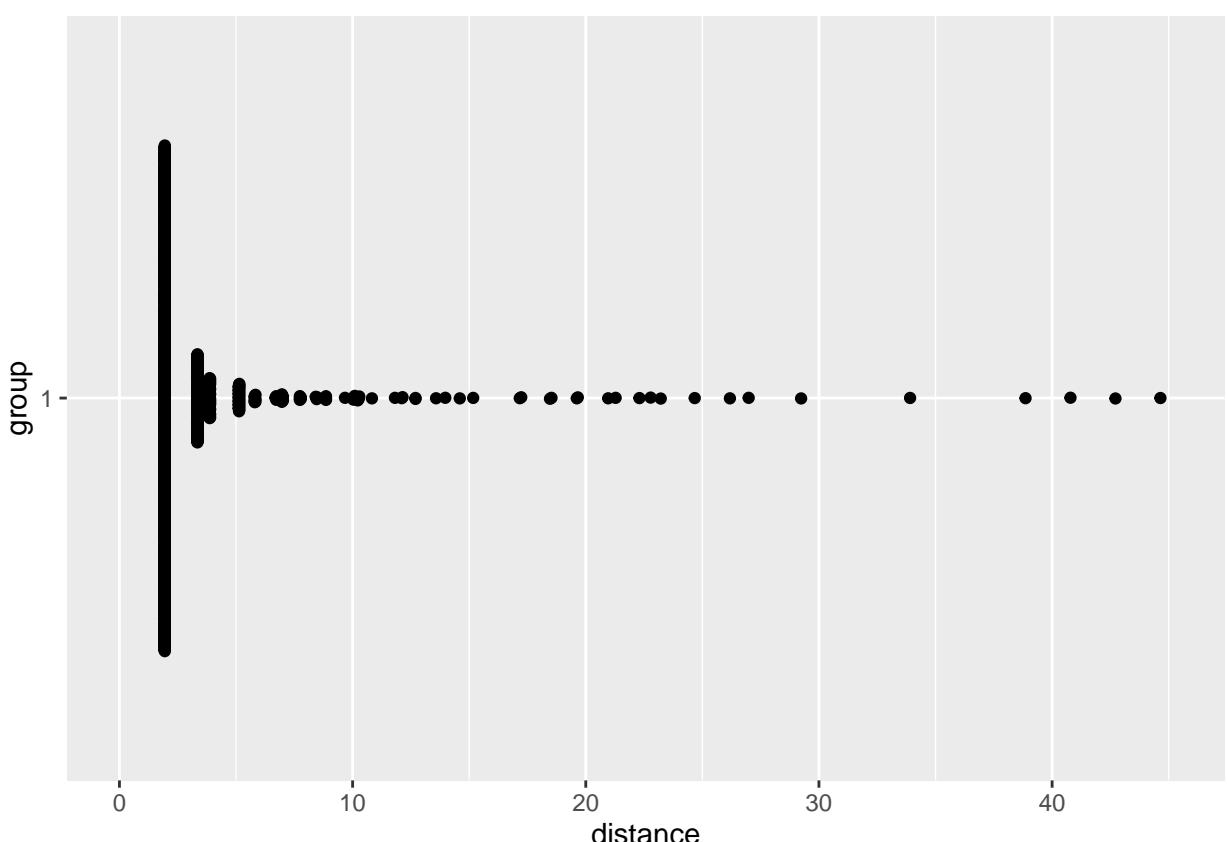
## To generate averaged high-D data

df_bin <- avg_highD_data(.data = df_all, column_start_text = "PC") ## Need to pass ID column name

## Compute 2D distances
distance <- cal_2d_dist(.data = tr_from_to_df)

plot_dist(distance)

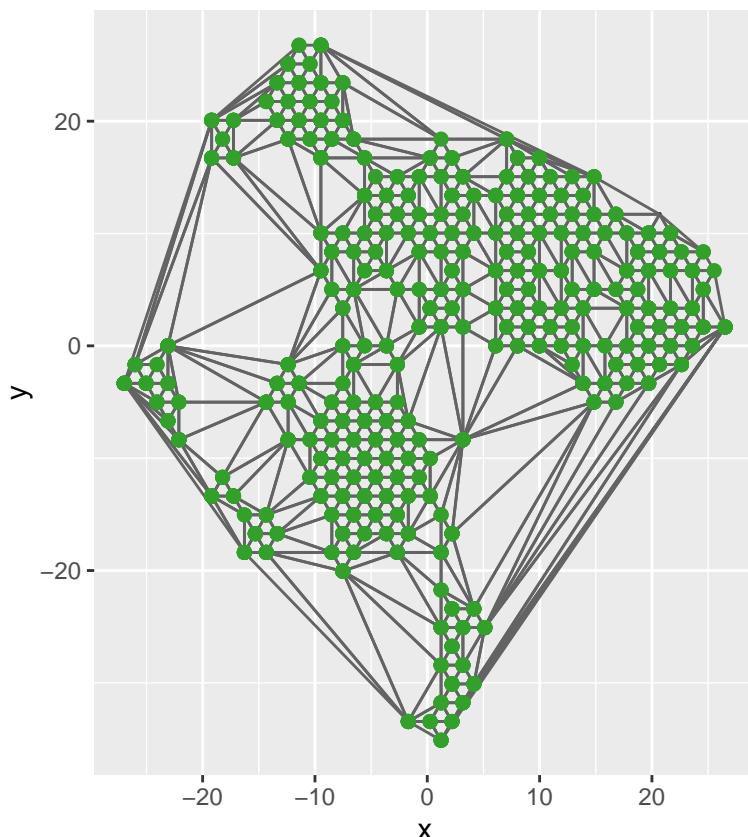
```



```
benchmark <- find_benchmark_value(.data = distance, distance_col = "distance")
```

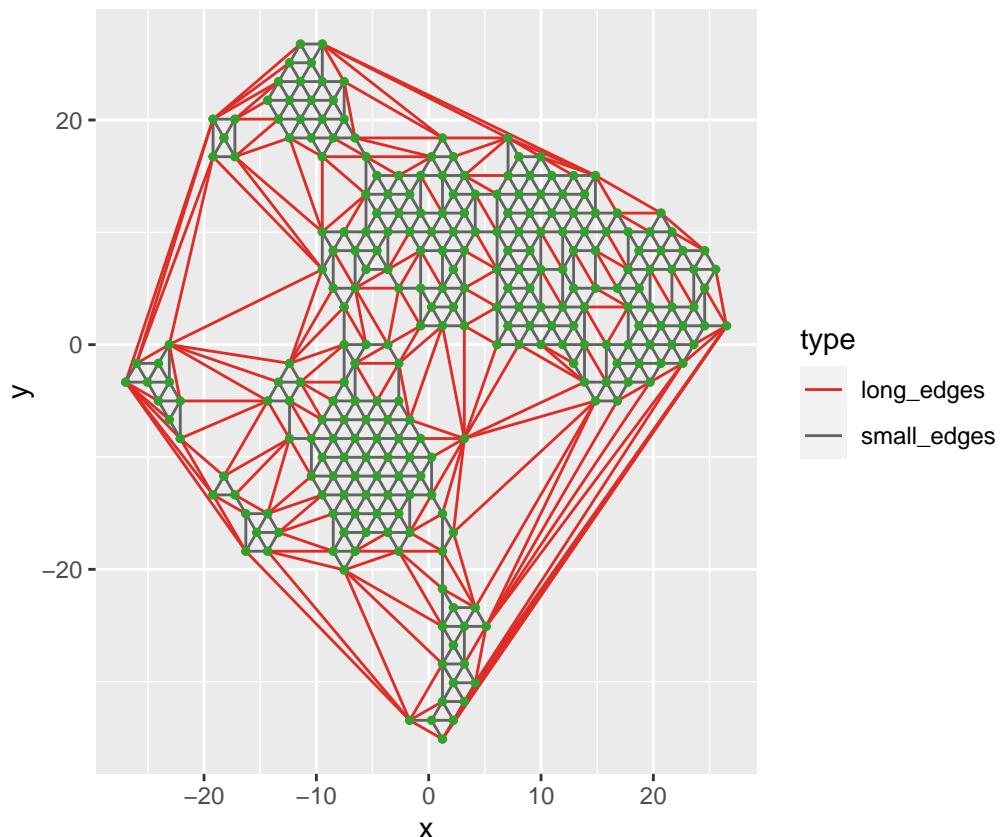
```
trimesh <- ggplot(df_bin_centroids, aes(x = x, y = y)) +  
  geom_point(size = 0.1) +  
  geom_trimesh() +  
  coord_equal()
```

```
trimesh
```

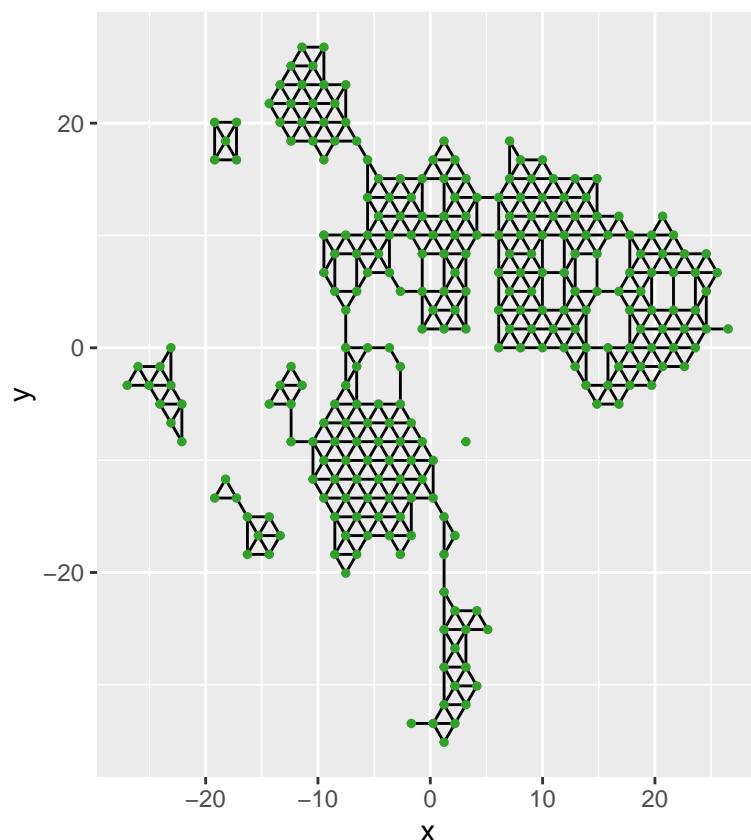


```
trimesh_gr <- colour_long_edges(.data = distance, benchmark_value = benchmark,  
                                triangular_object = tr1_object, distance_col = "distance")
```

```
trimesh_gr
```



```
trimesh_removed <- remove_long_edges(.data = distance, benchmark_value = benchmark,
                                      triangular_object = tr1_object, distance_col = "distance")
trimesh_removed
```



```
tour1 <- show_langevitour(df_all, df_bin, df_bin_centroids, benchmark_value = benchmark,
```

```
distance = distance, distance_col = "distance", column_start_text = "PC")  
tour1
```

4 Conclusion

5 Acknowledgements

This article is created using `knitr` (Xie 2015) and `rmarkdown` (Xie, Allaire, and Grolemund 2018) in R with the `rjtools::rjournal_article` template. The source code for reproducing this paper can be found at: <https://github.com/JayaniLakshika/paper-quollr>.

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