

732A98: Visualization - Cheatbook

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```
# Loading required R packages
library(ggplot2)
library(plotly)
library(shiny)
library(gridExtra)
library(xlsx)
library(MASS)
library(sf)
library(akima)
library(scales)
library(seriation)
library(dplyr)
library(crosstalk)
library(GGally)
library(tm)
library(wordcloud)
library(RColorBrewer)
library(htmltools)
library(tourr)
library(reshape)
library(ggraph)
library(igraph)
library(visNetwork)
library(data.table)
library(reshape2)
```

```
Sys.setenv('MAPBOX_TOKEN' = 'pk.eyJ1IjoibGFrc2hpZGFhIiwiYSI6ImNqbWIyOHN2NTRlZ3kzam10aT1jeGNybWgifQ.8EG9
```

Reading Data

Data Mugging

Quantile Computation

```
get_outliers <- function(x){
  quantile_values = quantile(x, probs = c(0.25, 0.75))
  q1 = quantile_values["25%"]
  q3 = quantile_values["75%"]

  return(c(which((x > (q3+1.5*(q3-q1)))), which(x < (q1-1.5*(q3-q1))))))
}
```

Scaling the Data

```
baseball_scaled <- scale(baseball_data[, 3:length(baseball_data)])
```

Distance Matrix between rows

```
distance_matrix <- dist(baseball_scaled, method = "euclidean")
```

Non-metric MDS

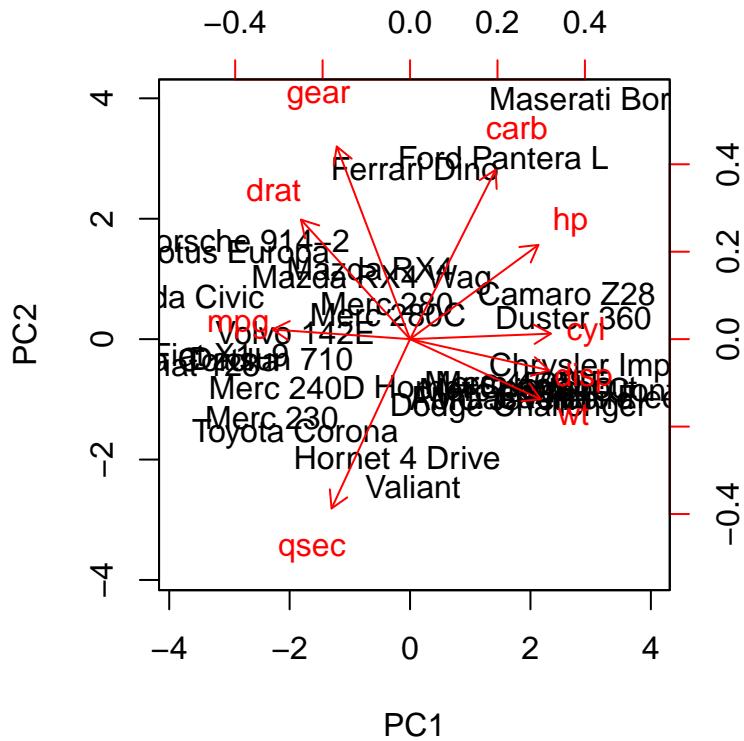
```
mds_result <- isoMDS(distance_matrix, k=2, p=2)

## initial value 19.856833
## iter 5 value 16.319153
## iter 10 value 16.046215
## final value 15.935476
## converged

coords <- mds_result$points
coords_mds <- as.data.frame(coords)
baseball_data_with_mds <- baseball_data
baseball_data_with_mds$MDS_V1 <- coords_mds$V1
baseball_data_with_mds$MDS_V2 <- coords_mds$V2
```

Principle Component Analysis

```
mtcars.pca <- prcomp(mtcars[, c(1:7, 10, 11)], center = TRUE, scale. = TRUE)
biplot(mtcars.pca, scale = 0)
```



Types of Projection

```
#projection = list(type = "mercator")
#projection = list(type = "albers usa")
#projection = list(type = "equirectangular")
#projection = list(type = "conic equal area")
#projection = list(type = "azimuthal equal area")
#projection = list(type = "equirectangular")
#projection = list(type = "orthographic")
```

Types of easing

```
#animation_opts(500, easing = 'linear', redraw = F)
# animation_opts(500, easing = 'quad', redraw = F)
# animation_opts(500, easing = 'cubic', redraw = F)
# animation_opts(500, easing = 'sin', redraw = F)
# animation_opts(500, easing = 'exp', redraw = F)
# animation_opts(500, easing = 'circle', redraw = F)
# animation_opts(500, easing = 'elastic', redraw = F)
# animation_opts(500, easing = 'back', redraw = F)
# animation_opts(500, easing = 'bounce', redraw = F)
# animation_opts(500, easing = 'linear-in', redraw = F)
# animation_opts(500, easing = 'quad-in', redraw = F)
```

```

# animation_opts(500, easing = 'cubic-in', redraw = F)
# animation_opts(500, easing = 'sin-in', redraw = F)
# animation_opts(500, easing = 'exp-in', redraw = F)
# animation_opts(500, easing = 'circle-in', redraw = F)
# animation_opts(500, easing = 'elastic-in', redraw = F)
# animation_opts(500, easing = 'back-in', redraw = F)
# animation_opts(500, easing = 'bounce-in', redraw = F)
# animation_opts(500, easing = 'linear-out', redraw = F)
# animation_opts(500, easing = 'quad-out', redraw = F)
# animation_opts(500, easing = 'cubic-out', redraw = F)
# animation_opts(500, easing = 'sin-out', redraw = F)
# animation_opts(500, easing = 'exp-out', redraw = F)
# animation_opts(500, easing = 'circle-out', redraw = F)
# animation_opts(500, easing = 'elastic-out', redraw = F)
# animation_opts(500, easing = 'back-out', redraw = F)
# animation_opts(500, easing = 'bounce-out', redraw = F)
# animation_opts(500, easing = 'linear-in-out', redraw = F)
# animation_opts(500, easing = 'quad-in-out', redraw = F)
# animation_opts(500, easing = 'cubic-in-out', redraw = F)
# animation_opts(500, easing = 'sin-in-out', redraw = F)
# animation_opts(500, easing = 'exp-in-out', redraw = F)
# animation_opts(500, easing = 'circle-in-out', redraw = F)
# animation_opts(500, easing = 'elastic-in-out', redraw = F)
# animation_opts(500, easing = 'back-in-out', redraw = F)
# animation_opts(500, easing = 'bounce-in-out', redraw = F)

```

Sorting dataset

```

mtcars <- as.data.table(mtcars)
mtcars <- mtcars[order(-mpg, cyl)]

head(mtcars, 10)

##      mpg cyl  disp  hp drat    wt  qsec vs am gear carb
## 1: 33.9   4  71.1  65 4.22 1.835 19.90  1  1     4     1
## 2: 32.4   4  78.7  66 4.08 2.200 19.47  1  1     4     1
## 3: 30.4   4  75.7  52 4.93 1.615 18.52  1  1     4     2
## 4: 30.4   4  95.1 113 3.77 1.513 16.90  1  1     5     2
## 5: 27.3   4  79.0  66 4.08 1.935 18.90  1  1     4     1
## 6: 26.0   4 120.3  91 4.43 2.140 16.70  0  1     5     2
## 7: 24.4   4 146.7  62 3.69 3.190 20.00  1  0     4     2
## 8: 22.8   4 108.0  93 3.85 2.320 18.61  1  1     4     1
## 9: 22.8   4 140.8  95 3.92 3.150 22.90  1  0     4     2
## 10: 21.5  4 120.1  97 3.70 2.465 20.01  1  0     3     1

```

Colour selection palette

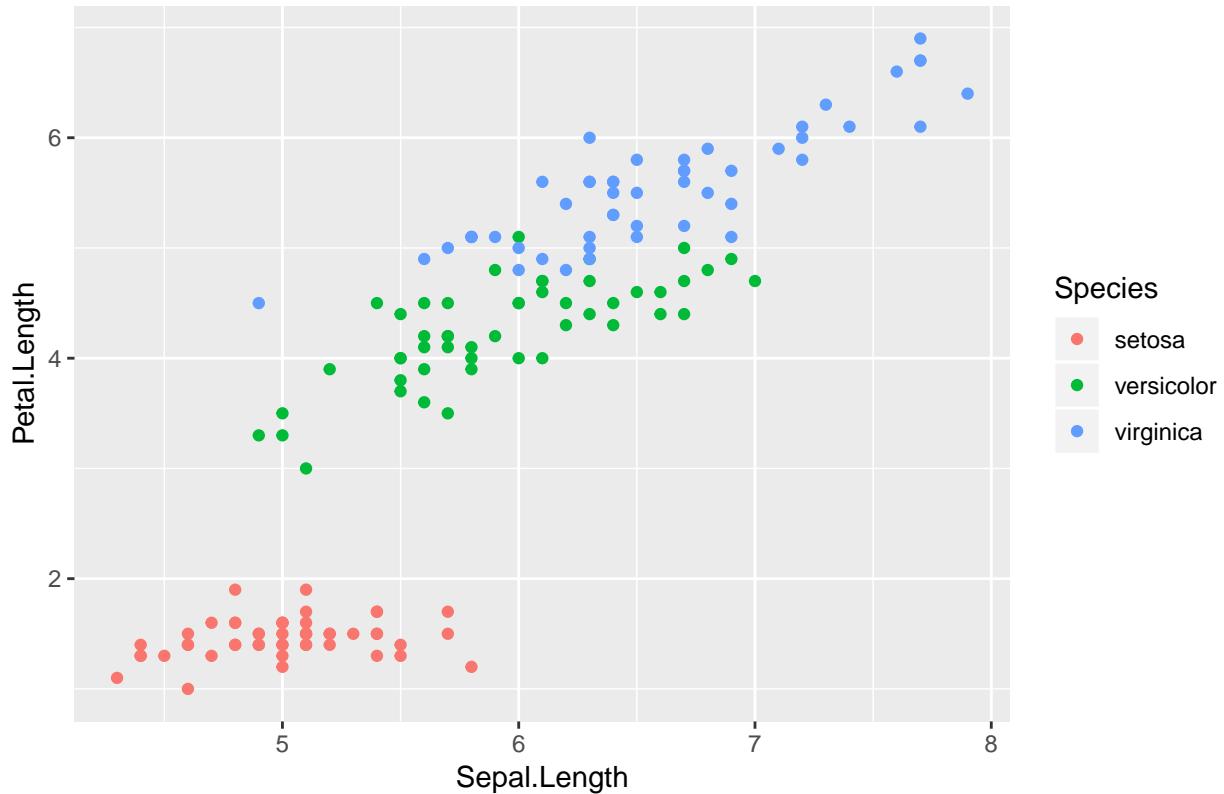
```

# The palette with grey:
default_colors <- c("#999999", "#E69F00", "#56B4E9", "#009E73", "#FOE442", "#0072B2", "#D55E00", "#CC79A0")

```

```
# ggplot2
ggplot(data=iris, aes(x= Sepal.Length, y = Petal.Length, color = Species)) + geom_point() + scale_fill_manual(values=c("red", "green", "blue"))
```

Custom Color for ggplot2

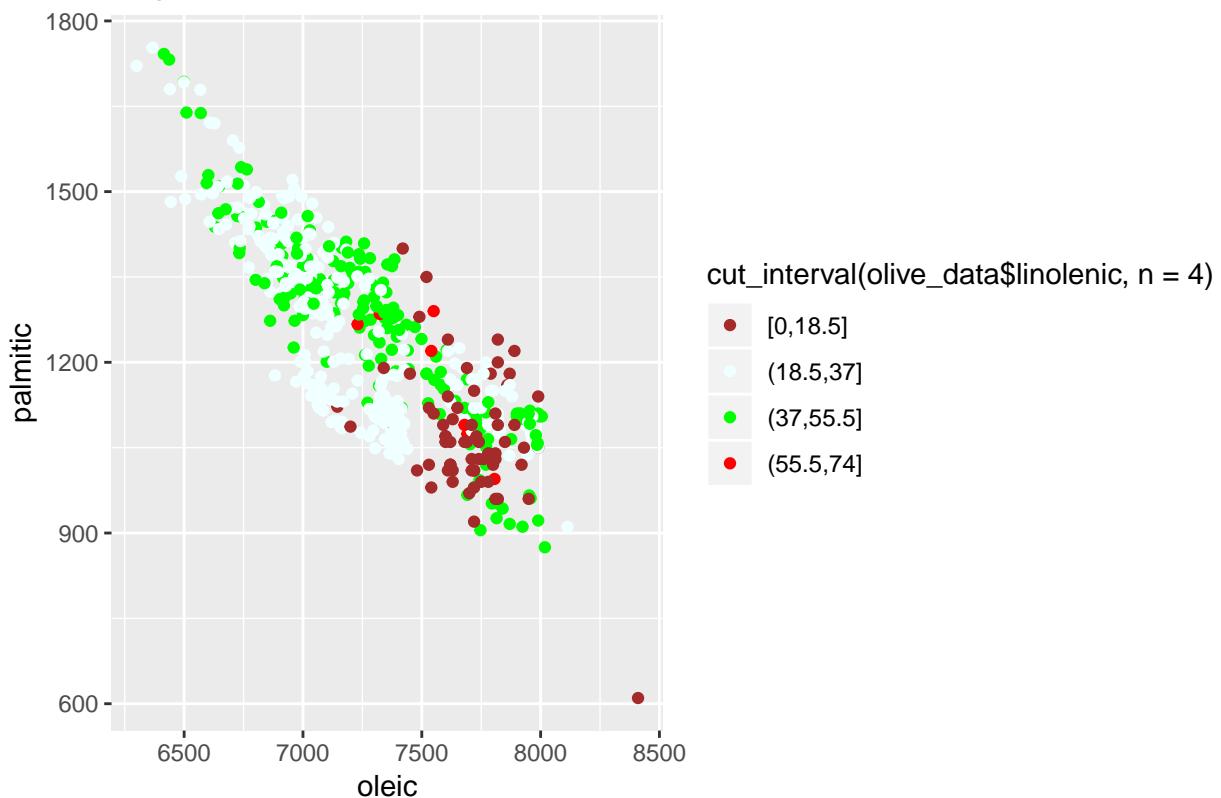


```
# plotly
x <- plot_ly(data = iris, x = ~Sepal.Length, y = ~Petal.Length, color = ~Species, colors = default_color)
```

Adding custom colours without palette

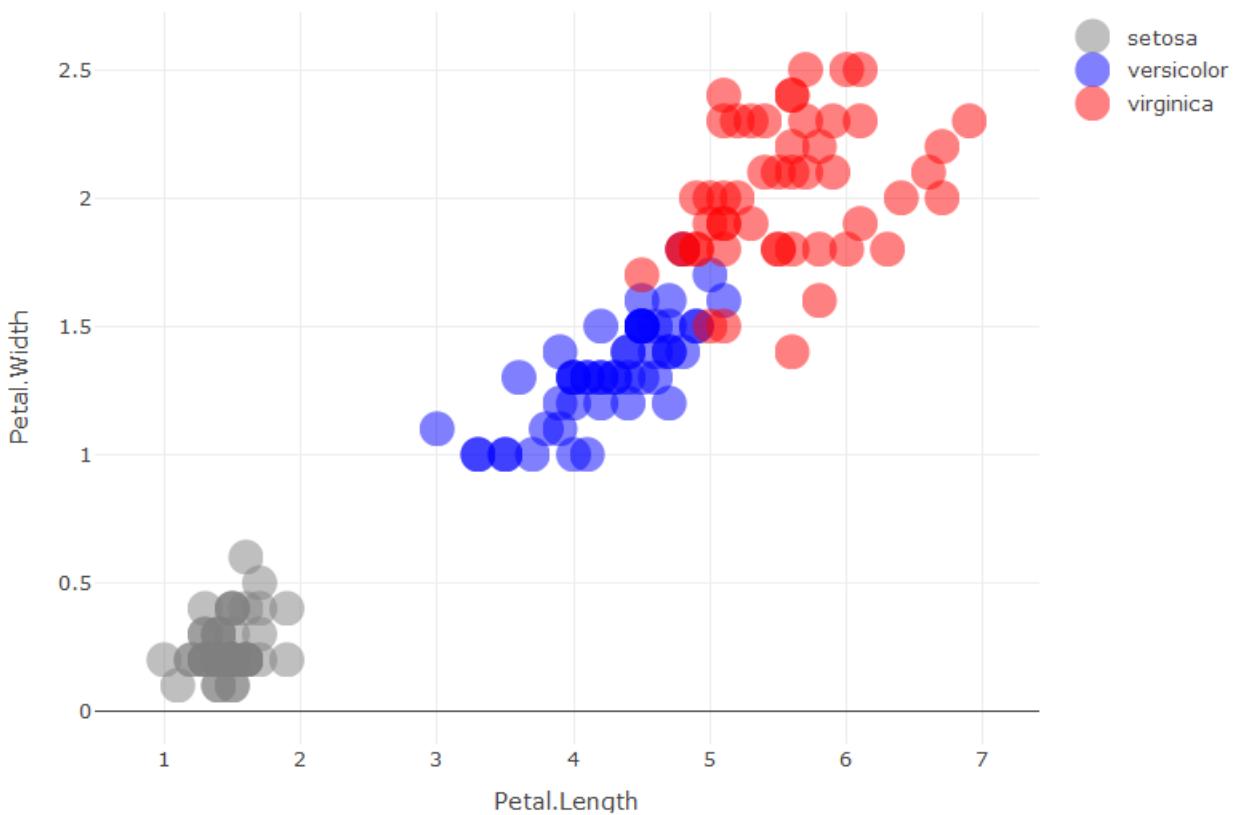
```
# ggplot2
ggplot(olive_data) +
  geom_point(aes(x = oleic, y = palmitic, color=cut_interval(olive_data$linolenic, n = 4))) +
  ggtitle("Dependence of Palmitic vs Oleic vs Linolenic") +
  scale_colour_manual(values=c("brown", "azure", "green", "red"))
```

Dependence of Palmitic vs Oleic vs Linolenic



```
# plotly
x <- plot_ly(iris, x = ~Petal.Length, y = ~Petal.Width,
              type="scatter", mode = "markers" , color = ~Species,
              colors = c("grey50", "blue", "red"), marker=list( size=20 , opacity=0.5))

knitr::include_graphics('./plotly_custom_color.png')
```



“

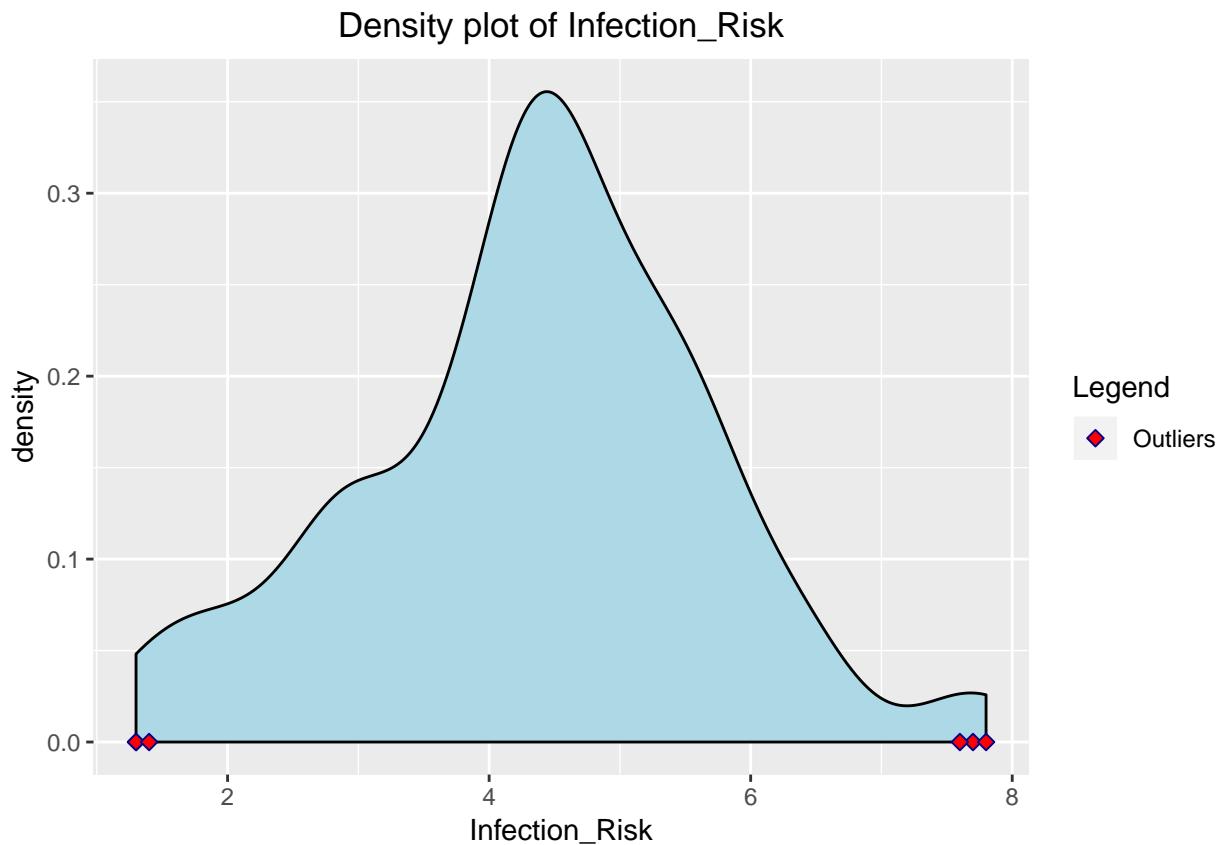
Single Plots

Density Plot

Density Plot with Outlier Highlight using GGplot2

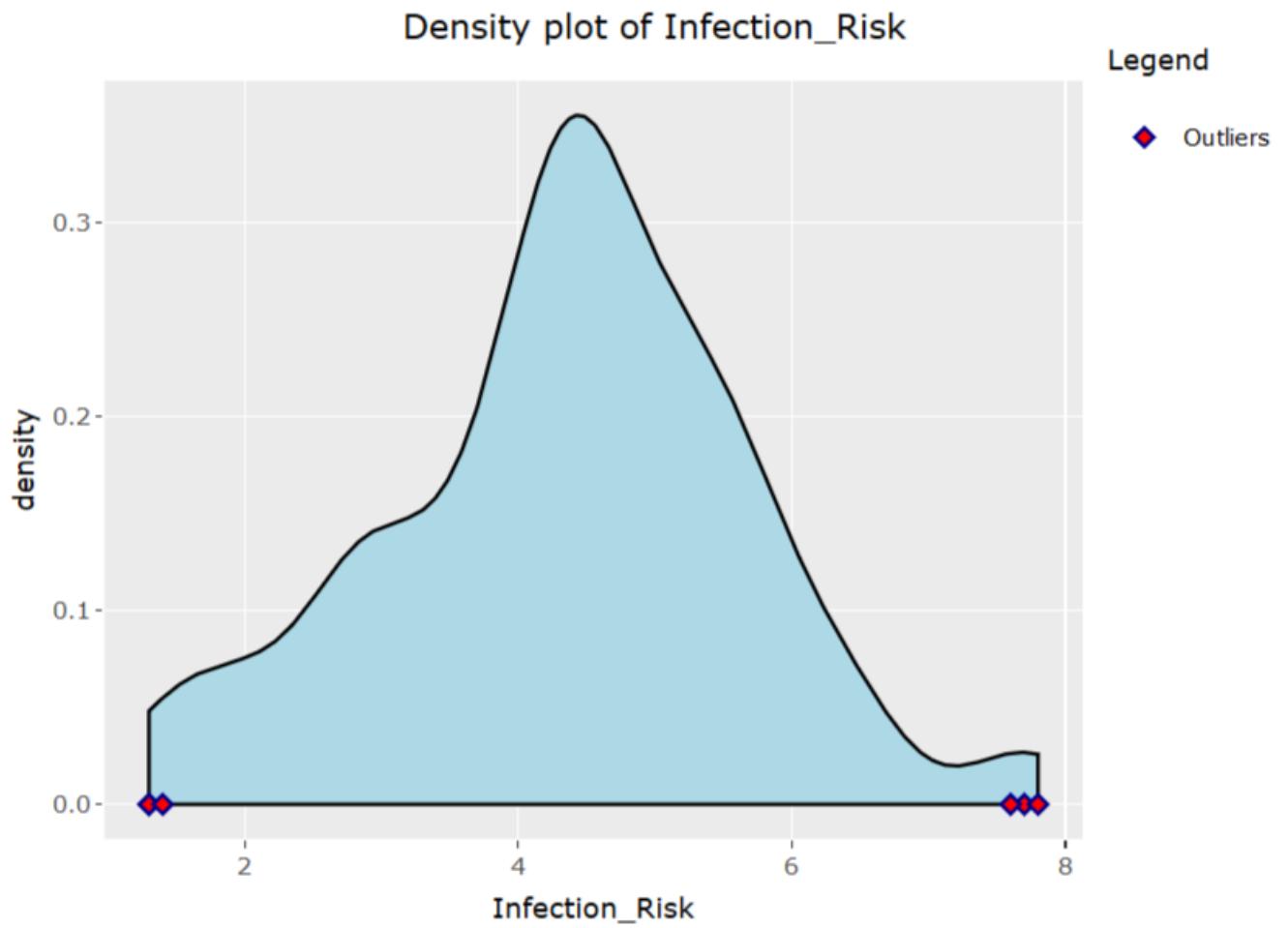
```
density_plot_infection_risk = ggplot(senic_data) +
  ggtitle("Density plot of Infection_Risk") +
  geom_density(aes(x=Infection_Risk), fill = "lightblue") +
  geom_point(data=senic_data[get_outliers(senic_data$Infection_Risk),],
             aes(x=Infection_Risk, y=0, colour="Outliers"),
             shape=23, size=2, fill="red") +
  scale_color_manual(values = c("darkblue","black")) +
  labs(colour="Legend") +
  theme(plot.title = element_text(hjust = 0.5), legend.position = "right")

density_plot_infection_risk
```



Density Plot with Outlier Highlight using Plotly (converting from ggplot2)

```
x <- ggplotly(p=density_plot_infection_risk)
knitr:::include_graphics('./3.1.2.png')
```



Histogram Plot

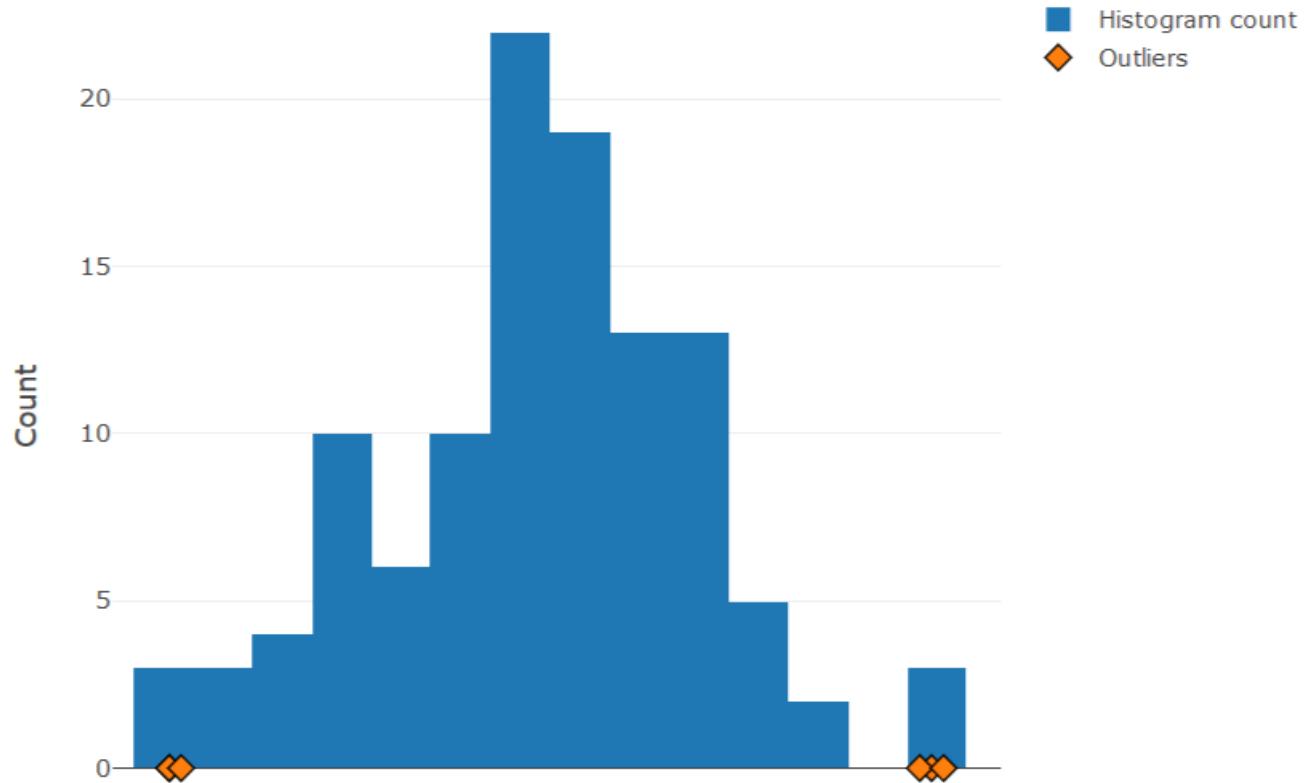
Histogram plot with Outlier Highlight using Plotly

```
outliers = senic_data[get_outliers(senic_data$Infection_Risk),c("Infection_Risk")]
senic_data$zero = 0

x <- plot_ly(senic_data, x=~Infection_Risk) %>%
  add_histogram(name="Histogram count") %>%
  filter(is.element(Infection_Risk, outliers)) %>%
  add_markers(x=~Infection_Risk,y=~zero, name="Outliers",
              marker=list(symbol="diamond", size=10, line = list(color="black", width=1))) %>%
  layout(title="Histogram of Infection_Risk", yaxis=list(title="Count"))

knitr:::include_graphics('./3.2.1.png')
```

Histogram of Infection_Risk

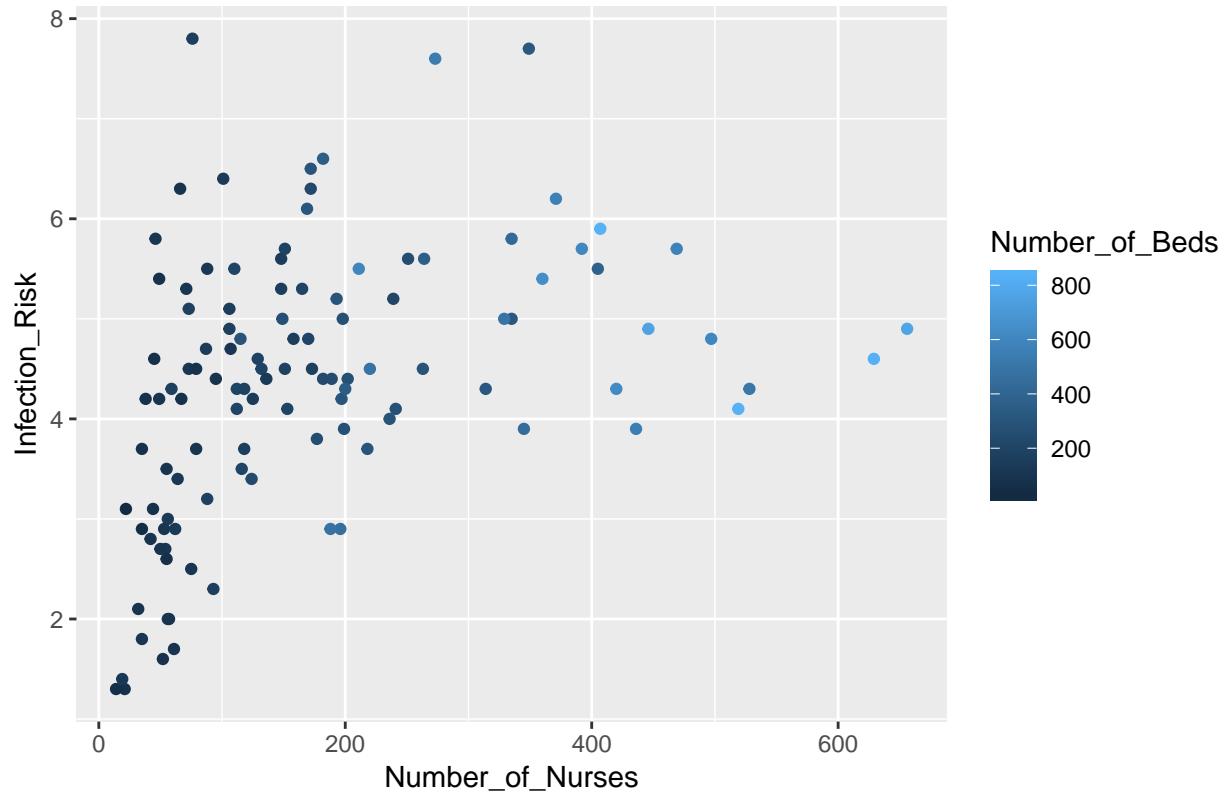


Scatter Plot

Simple scatter plot with colour

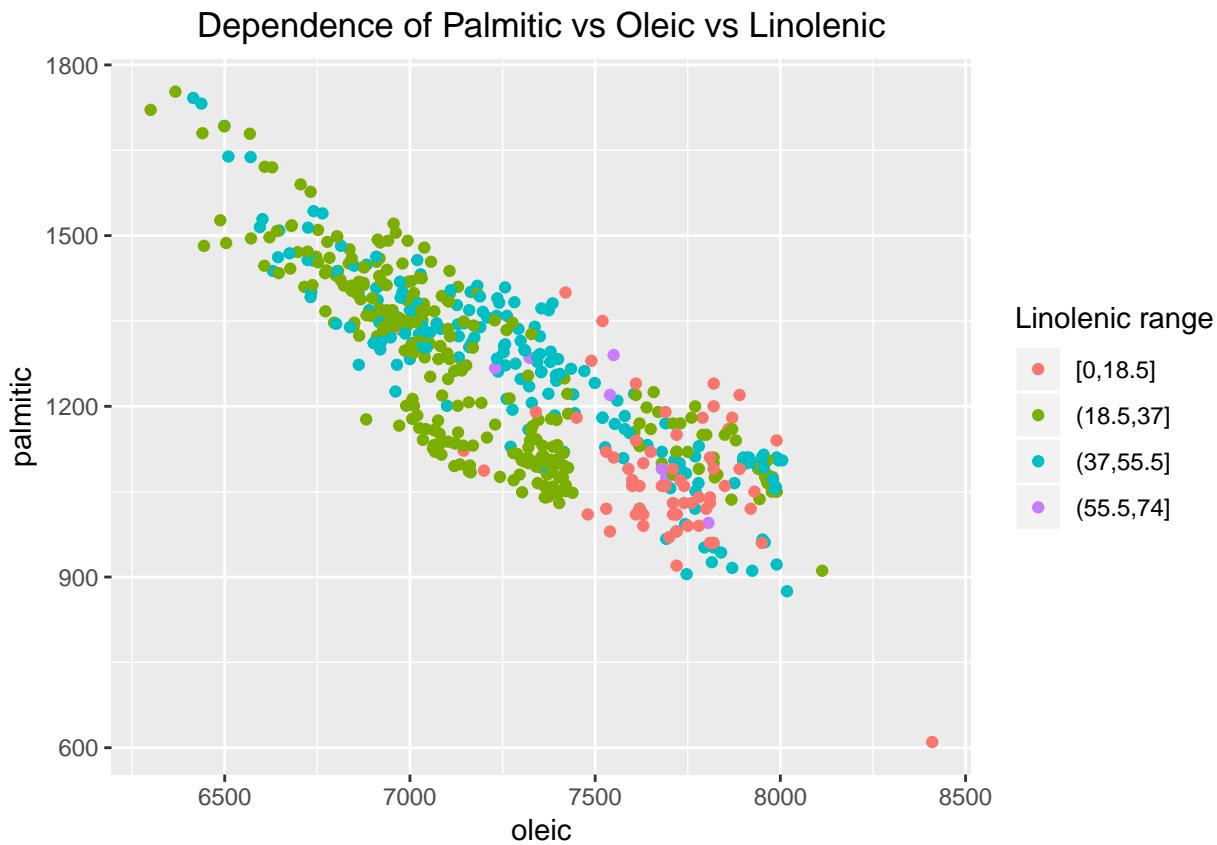
```
ggplot(senic_data) + geom_point(aes(x=Number_of_Nurses, y=Infection_Risk, color=Number_of_Beds)) +
  ggtitle("Scatterplot of Infection_Risk vs Number_of_Nurses") +
  theme(plot.title = element_text(hjust = 0.5))
```

Scatterplot of Infection_Risk vs Number_of_Nurses



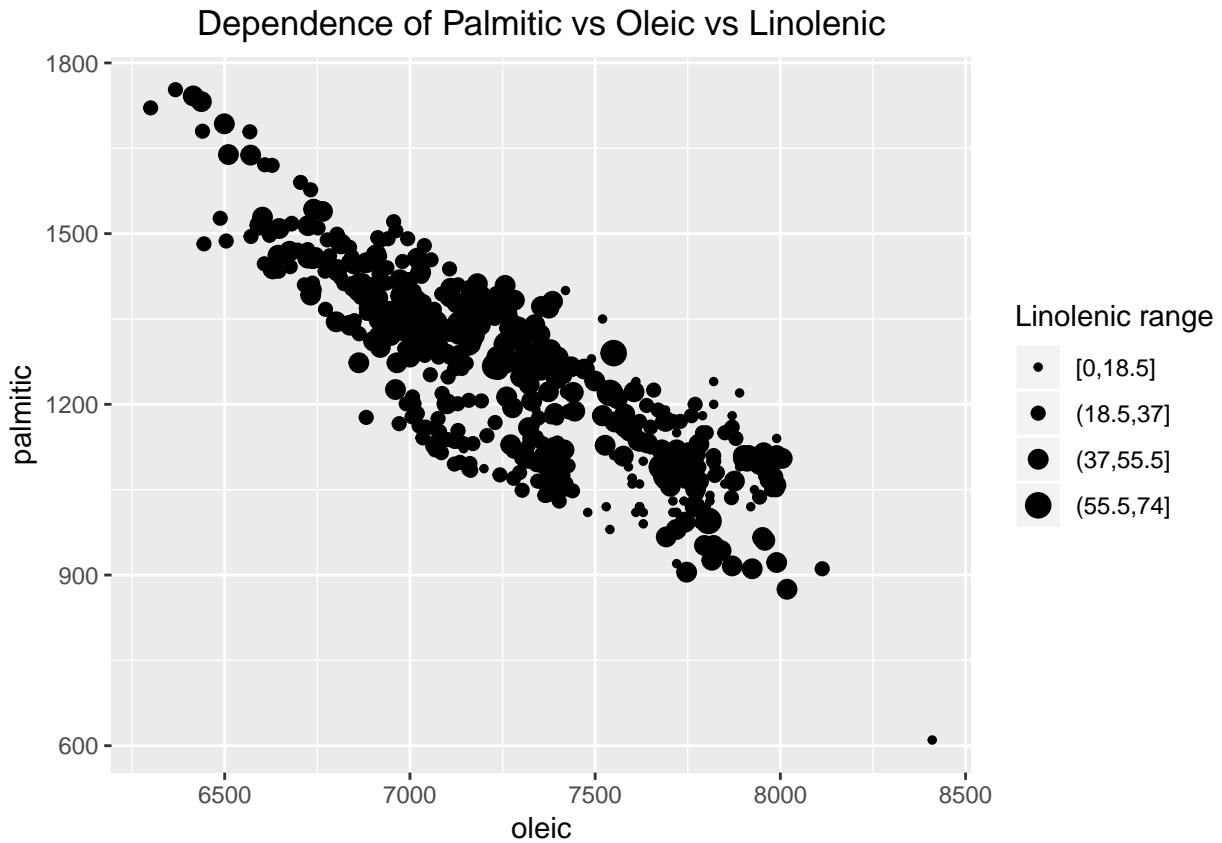
Scatter Plot with Discretization (split a variable into classes)

```
ggplot(olive_data) +
  geom_point(aes(x = oleic, y = palmitic,
                 color=cut_interval(olive_data$linolenic, n = 4))) +
  ggtitle("Dependence of Palmitic vs Oleic vs Linolenic") +
  theme(plot.title = element_text(hjust = 0.5)) +
  labs(color = 'Linolenic range')
```



Scatter plot size varied

```
ggplot(olive_data) + geom_point(aes(x = oleic, y = palmitic, size = cut_interval(linolenic, n = 4))) +
  ggtitle("Dependence of Palmitic vs Oleic vs Linolenic") +
  theme(plot.title = element_text(hjust = 0.5)) +
  scale_size_manual(name = "Linolenic range", values = c(1, 2, 3, 4))
```

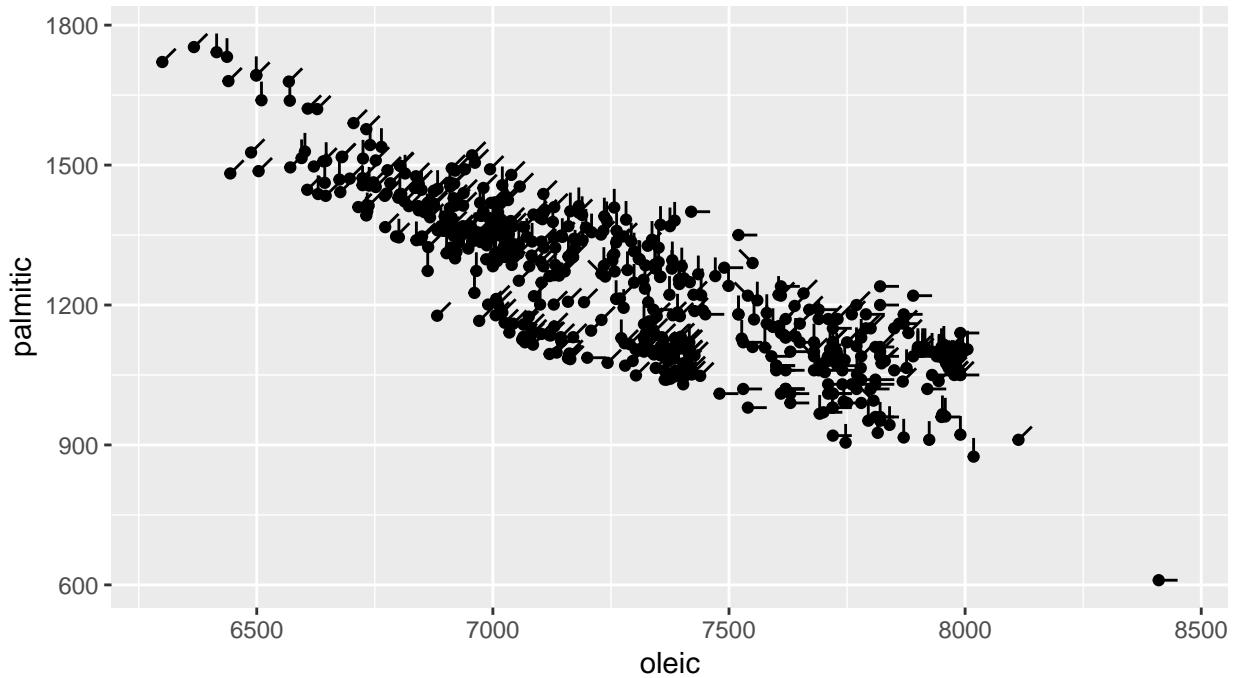


Scatter plot angle varied

```
# Pre-processing - Setting angle values based on category
olive_data$linolenic_class <- cut_interval(olive_data$linolenic, n = 4)
levels(olive_data$linolenic_class) <- (0:3) * (pi/4)
olive_data$linolenic_class <- as.numeric(as.character(olive_data$linolenic_class))

ggplot(olive_data, aes(x=oleic, y=palmitic)) + geom_point() +
  geom_spoke(aes(angle = olive_data$linolenic_class), radius=40) +
  ggtitle("Dependence of Palmitic vs Oleic vs Linolenic
Legend
Orientation angle of spoke : Linolenic class
0:(0,18.5], 45:(18.5,37], 90:(37,55.5], 135:(0,18.5] ") +
  theme(plot.title = element_text(hjust = 0.5))
```

Dependence of Palmitic vs Oleic vs Linolenic
 Legend
 Orientation angle of spoke : Linolenic class
 $0:(0,18.5]$, $45:(18.5,37]$, $90:(37,55.5]$, $135:(0,18.5]$



Scatter plot with color, size and shape

```
ggplot(olive_data)+  

  geom_point(aes(x = oleic, y = eicosenoic, color = cut_interval(linoleic, n = 3),  

                 shape = cut_interval(palmitic, n = 3),  

                 size = cut_interval(palmitoleic, n = 3))) +  

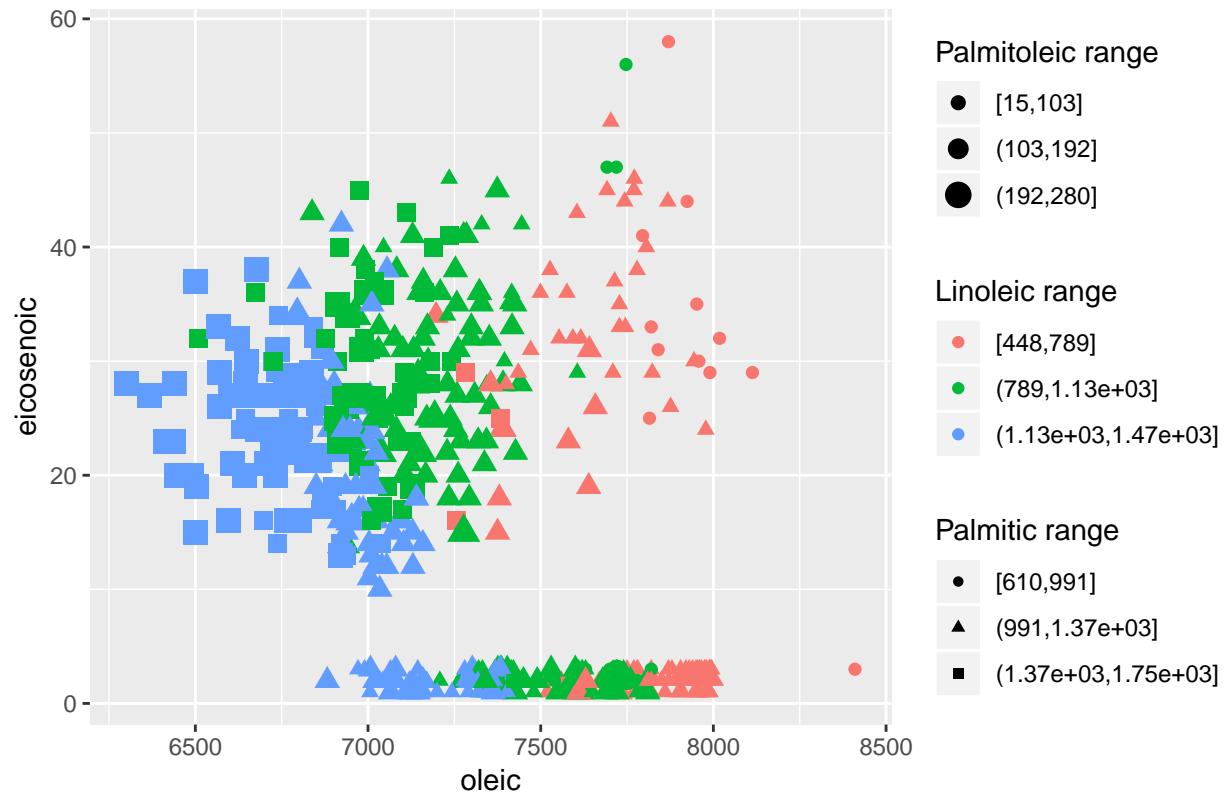
  scale_size_manual(values = c(2,3,4)) +  

  labs(shape = "Palmitic range", color = "Linoleic range", size = "Palmitoleic range") +  

  ggtitle("Dependence of Oleic vs Eicosenoic vs Linoleic vs Palmitic vs Palmitoleic") +  

  theme(plot.title = element_text(hjust = 0.5))
```

endence of Oleic vs Eicosenoic vs Linoleic vs Palmitic vs Palmitoleic



Scatter Plot of MDS

```
baseball_scaled <- scale(baseball_data[, 3:length(baseball_data)])

### Distance Matrix between rows
distance_matrix <- dist(baseball_scaled, method = "euclidean")

### Non-metric MDS
mds_result <- isoMDS(distance_matrix, k=2, p=2)

## initial value 19.856833
## iter 5 value 16.319153
## iter 10 value 16.046215
## final value 15.935476
## converged

coords <- mds_result$points
coords_mds <- as.data.frame(coords)
baseball_data_with_mds <- baseball_data
baseball_data_with_mds$MDS_V1 <- coords_mds$V1
baseball_data_with_mds$MDS_V2 <- coords_mds$V2

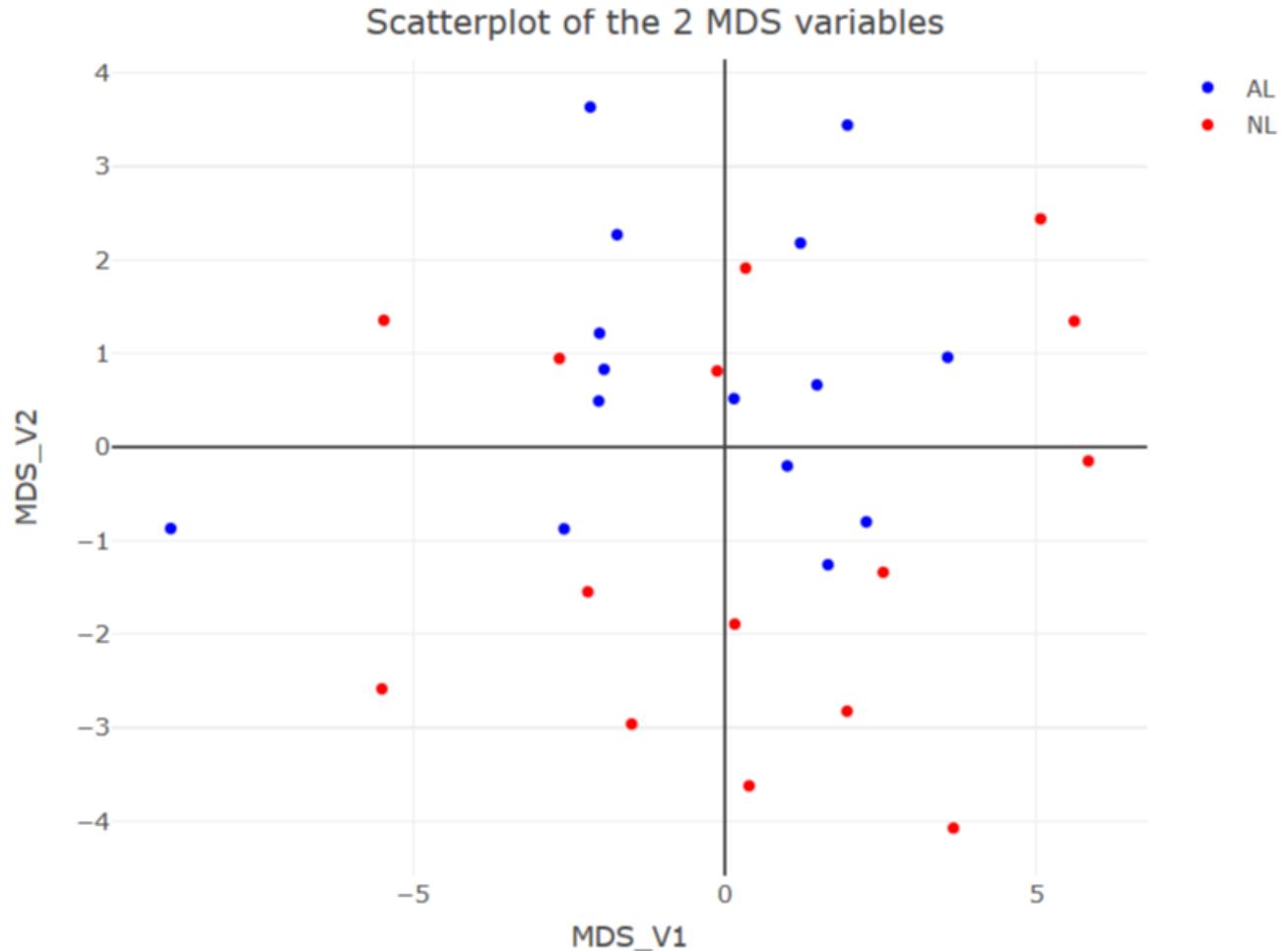
x <- plot_ly(baseball_data_with_mds, x=~MDS_V1, y=~MDS_V2) %>%
```

```

add_markers(color=~League, colors = c("blue", "red"),
            text=baseball_data_with_mds$Team, hoverinfo="text") %>%
layout(title="Scatterplot of the 2 MDS variables")

knitr::include_graphics('~/3.3.6.png')

```



Shepard Plot

Shepard plot shows the fit of MDS, the distance in original dataset vs. distance in reordered dataset should be similar

```

baseball_scaled <- scale(baseball_data[,3:length(baseball_data)])

### Distance Matrix between rows
distance_matrix <- dist(baseball_scaled, method = "euclidean")
mds_result <- isoMDS(distance_matrix, k=2, p=2)

## initial value 19.856833
## iter 5 value 16.319153
## iter 10 value 16.046215
## final value 15.935476

```

```

## converged
coords <- mds_result$points

shp <- Shepard(distance_matrix, coords)
delta <- as.numeric(distance_matrix)
D <- as.numeric(dist(coords, method = "euclidean"))

n <- nrow(coords)
index <- matrix(1:n, nrow=n, ncol=n)
index1 <- as.numeric(index[lower.tri(index)])

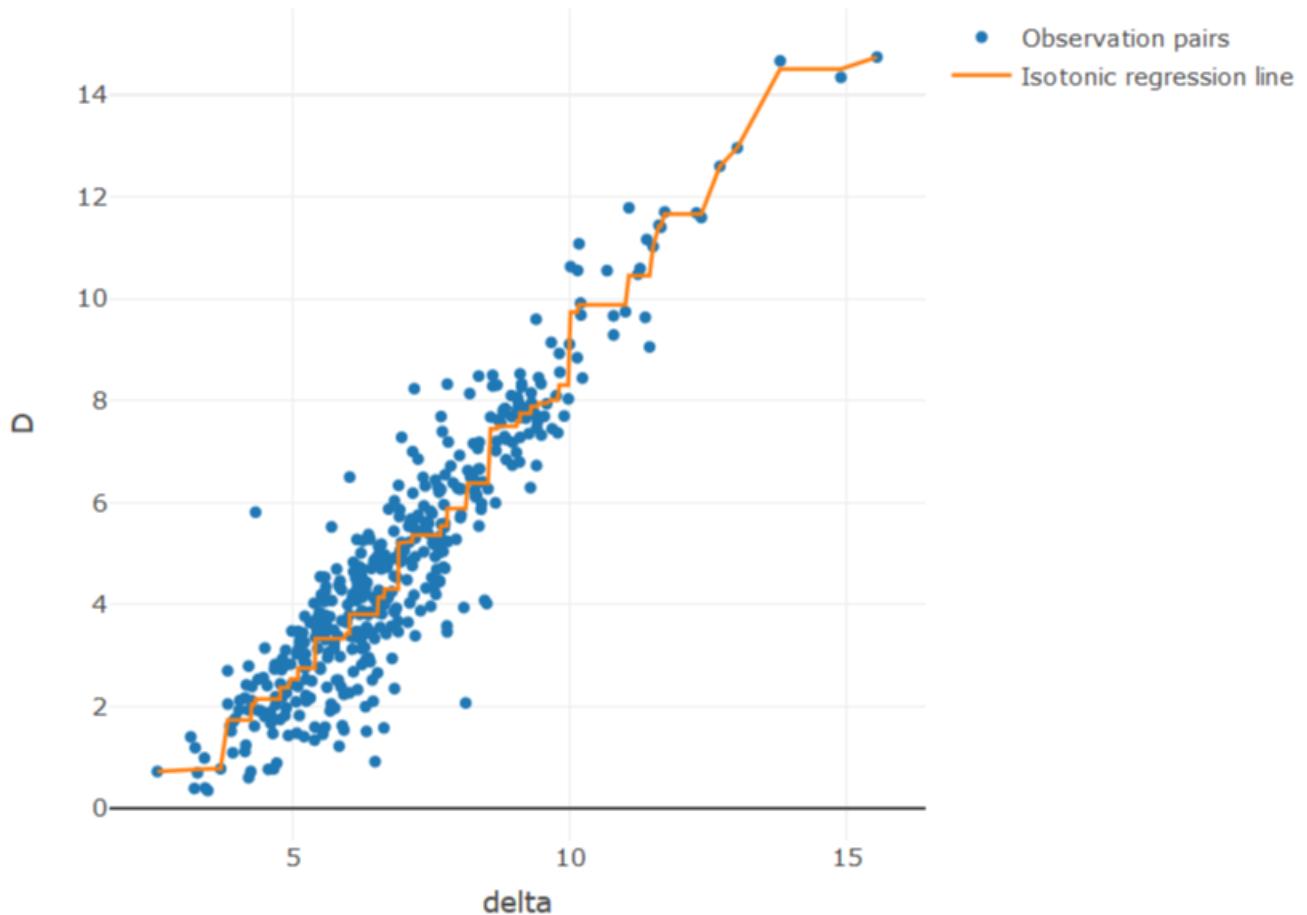
n <- nrow(coords)
index <- matrix(1:n, nrow=n, ncol=n, byrow = T)
index2 <- as.numeric(index[lower.tri(index)])

x <- plot_ly()%>%
  add_markers(x=-delta, y=-D, name="Observation pairs", hoverinfo = 'text',
              text = ~paste('Obj 1: ',
                            rownames(baseball_data_with_mds)[index1],
                            '<br> Obj 2: ',
                            rownames(baseball_data_with_mds)[index2])) %>%
  add_lines(x=-shp$x, y=-shp$yf, name="Isotonic regression line") %>%
  layout(title="Shepard's plot of MDS operation")

knitr::include_graphics('./3.4.1.png')

```

Shepard's plot of MDS operation



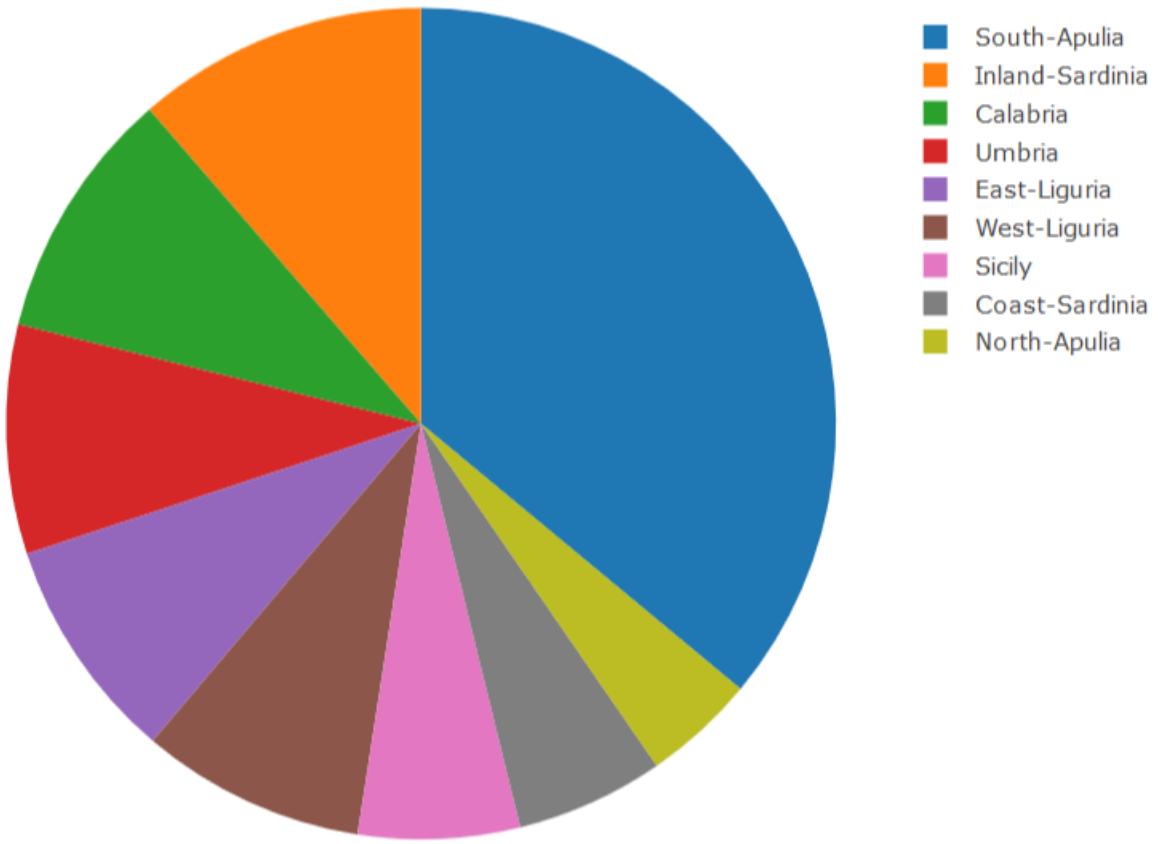
Pie Charts

Simple pie chart

```
x <- plot_ly(olive_data, labels=~Area, type='pie', textinfo = "none") %>%
  layout(title = "Pie chart of proportion of oils coming from different areas")

knitr::include_graphics('./3.5.1.png')
```

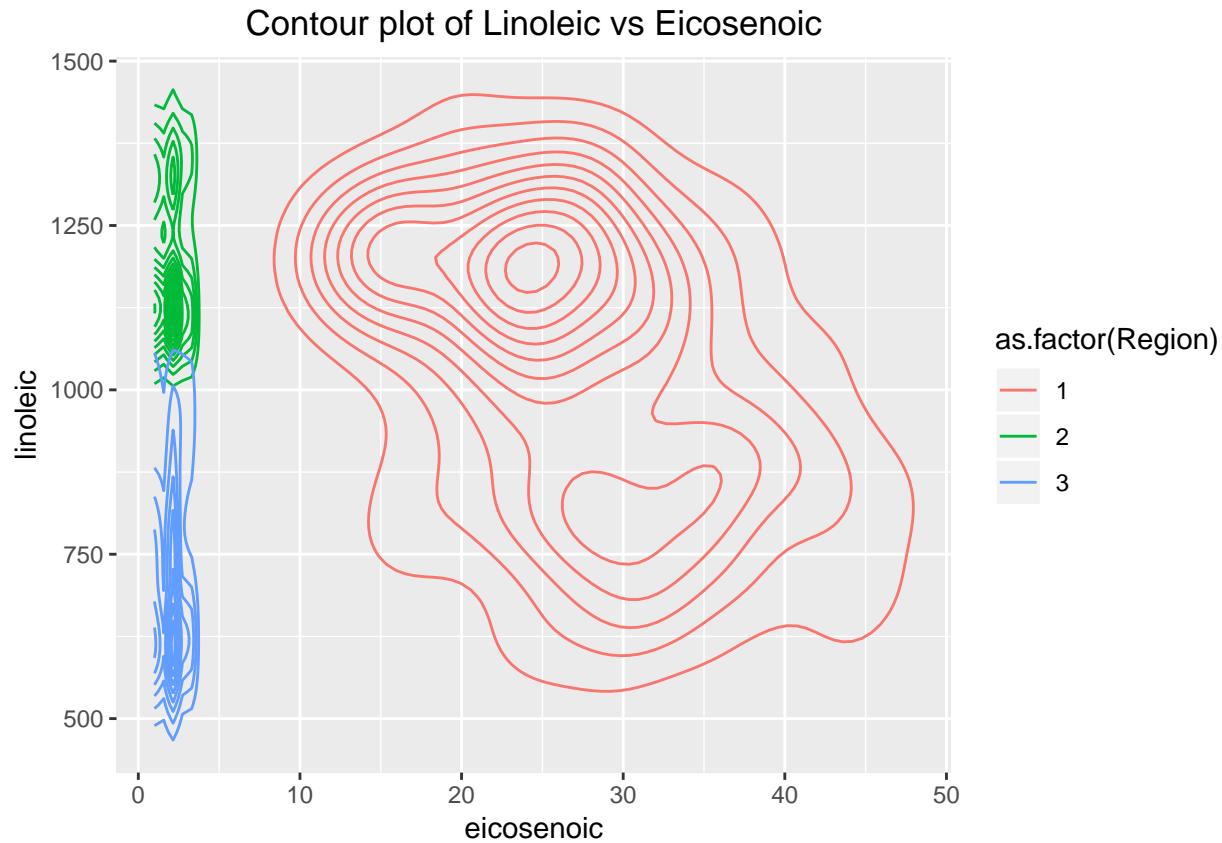
Pie chart of proportion of oils coming from different areas



2D density contour plot

Simple 2D plot using ggplot2

```
ggplot(olive_data)+geom_density_2d(aes(x=eicosenoic, y=linoleic, colour=as.factor(Region)))+  
  ggtitle("Contour plot of Linoleic vs Eicosenoic") +  
  theme(plot.title = element_text(hjust = 0.5))
```



Dot Map Plots (World Map)

Dot Map (Map with scatter plots)

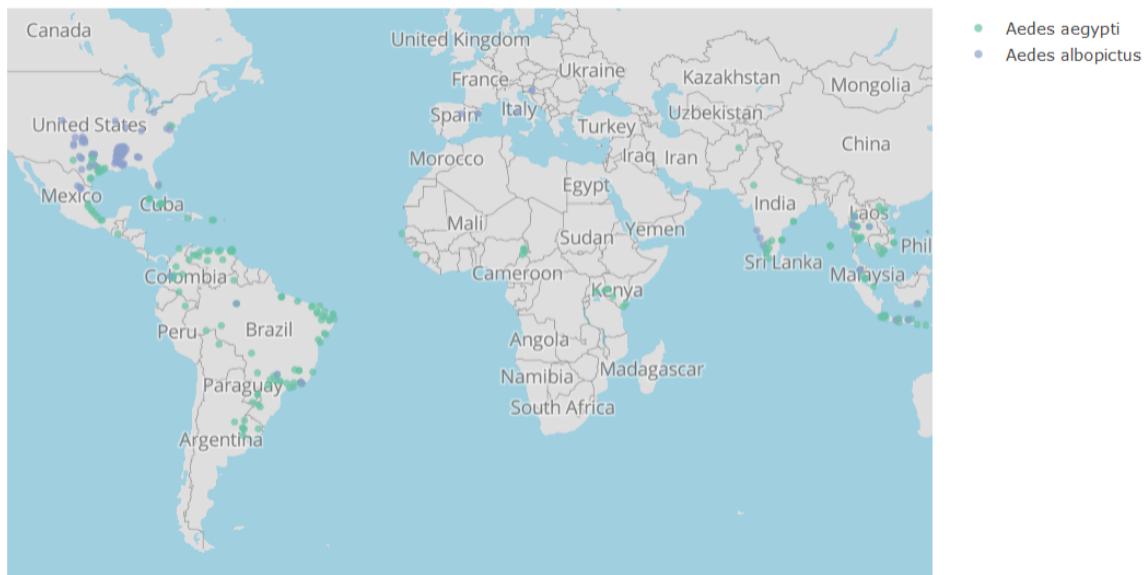
```

x <- plot_mapbox(aegypti_data[aegypti_data$YEAR == 2004], lat = ~Y, lon = ~X) %>%
  add_markers(color = ~VECTOR, hoverinfo = "text",
             text = ~paste(COUNTRY), alpha = 0.7) %>%
  layout(title = "Dot map of mosquito distribution in the world (2004)")

knitr:::include_graphics('./3.7.1.png')

```

Dot map of mosquito distribution in the world (2004)



Draw line between two places

Plotly between two places

```
x <- plot_geo(lat = c(40.7127, 51.5072), lon = c(-74.0059, 0.1275)) %>%
  add_lines(color = I("blue"), size = I(2)) %>%
  layout(
    title = 'London to NYC Great Circle',
    showlegend = FALSE,
    geo = list(
      resolution = 50,
      showland = TRUE,
      showlakes = TRUE,
      landcolor = toRGB("grey80"),
      countrycolor = toRGB("grey80"),
      lakecolor = toRGB("white"),
      projection = list(type = "equirectangular"),
      coastlinewidth = 2,
      lataxis = list(
        range = c(20, 60),
        showgrid = TRUE,
        tickmode = "linear",
        dtick = 10
      ),
      lonaxis = list(
        range = c(-100, 20),
        showgrid = TRUE,
        tickmode = "linear",
        dtick = 20
      )
    )
  )
```

```

)
knitr:::include_graphics('./world_map.png')

```

London to NYC Great Circle



Choropleth Map

Choropleth Map with Equirectangular Projection

```

# Data aggregation
country_aggregate = aggregate(aegypti_data[,c("COUNTRY", "COUNTRY_ID")],
                               by = list(aegypti_data$COUNTRY, aegypti_data$COUNTRY_ID), FUN=length)
country_aggregate$COUNTRY = NULL
colnames(country_aggregate) = c("COUNTRY", "COUNTRY_ID", "Count")

x <- plot_geo(country_aggregate) %>%
  add_trace(
    z = ~Count,
    text = ~COUNTRY, locations = ~COUNTRY_ID) %>%
  layout(title = "Choropleth plot of number of mosquitoes",
         geo = list(projection = list(type = "equirectangular")))

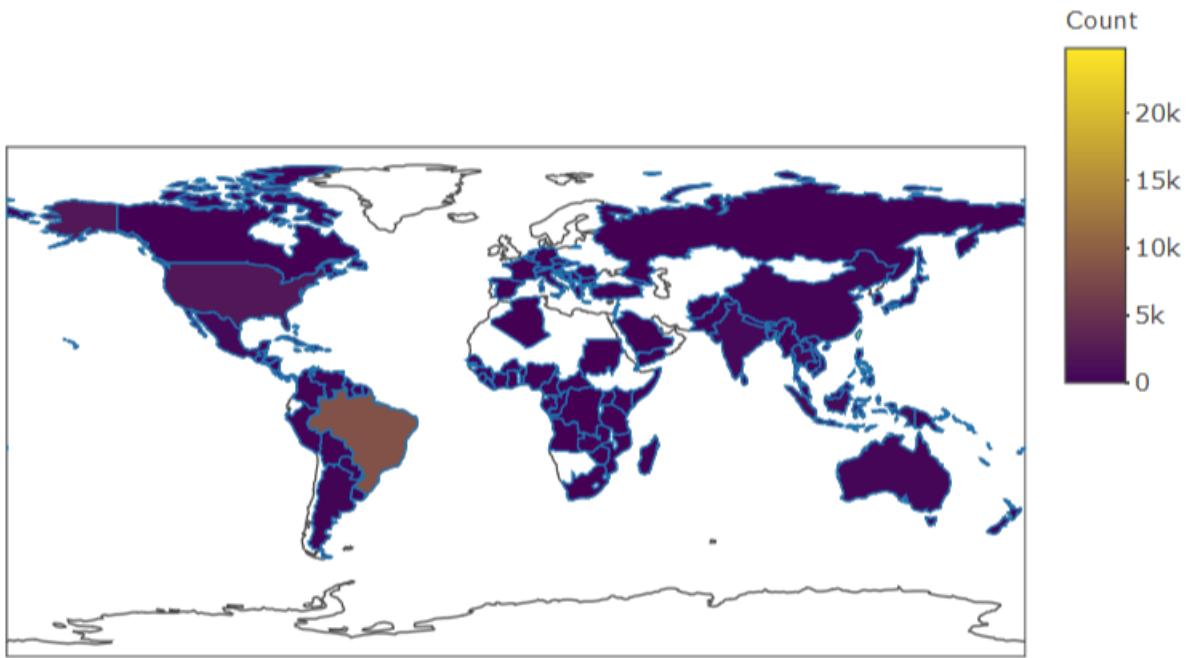
```

```

knitr:::include_graphics('./3.8.1.png')

```

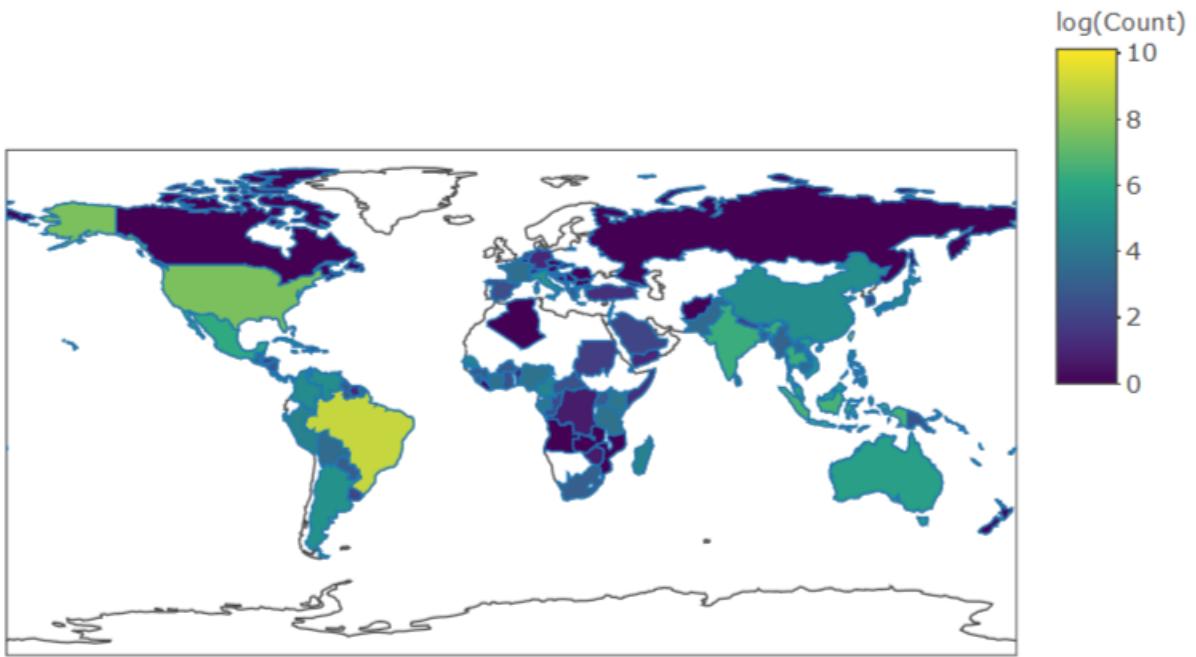
Choropleth plot of number of mosquitoes



Choropleth plot with Equirectangular Projection and log

```
x <- plot_geo(country_aggregate) %>%
  add_trace(
    z = -log(Count) ,
    text = -paste(COUNTRY, "\n Count: ", Count), locations = ~COUNTRY_ID,
    hoverinfo = "text"
  ) %>%
  layout(
    title = "Choropleth plot of number of mosquitoes",
    geo = list(projection = list(type = "equirectangular")))
knitr:::include_graphics('./3.8.2.png')
```

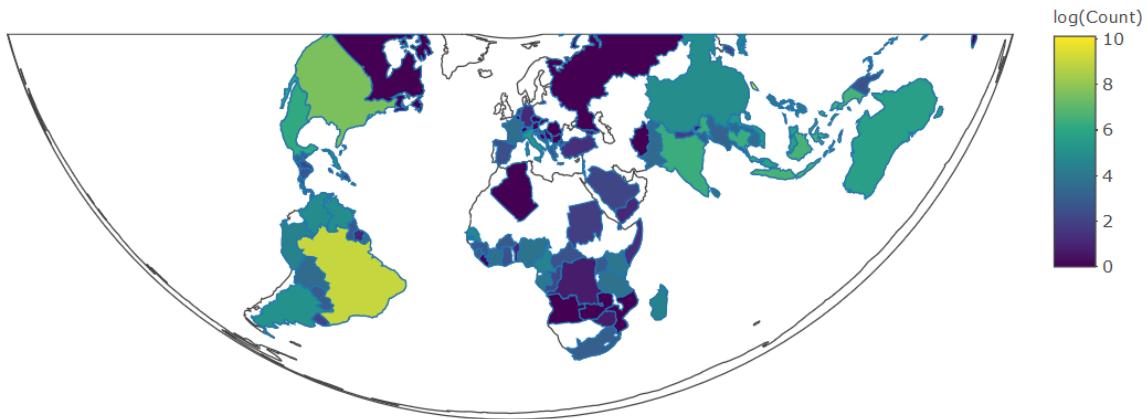
Choropleth plot of number of mosquitoes



Choropleth plot with Conic Area Projection and log

```
x <- plot_geo(country_aggregate) %>%
  add_trace(
    z = ~log(Count) ,
    text = ~paste(COUNTRY, "\n Count: ", Count), locations = ~COUNTRY_ID
  ) %>%
  layout(
    title = "Choropleth plot of number of mosquitoes",
    geo = list(projection = list(type = "conic equal area")))
knitr:::include_graphics('3.8.3.png')
```

Choropleth plot of number of mosquitoes



Choropleth plot using custom shape files(sf file)

```
swe_data = read.csv("000000KD.csv")

swe_data_processed = data.frame(region = unique(swe_data$region))
swe_data_split = split(swe_data, swe_data$age)
for (i in seq_along(swe_data_split)) {
  swe_data_processed[[names(swe_data_split)[i]]] = merge(swe_data_split[[i]],
                                                       swe_data_processed$region,
                                                       by.x = 'region',
                                                       by.y = 1, all = T)$X2016
}

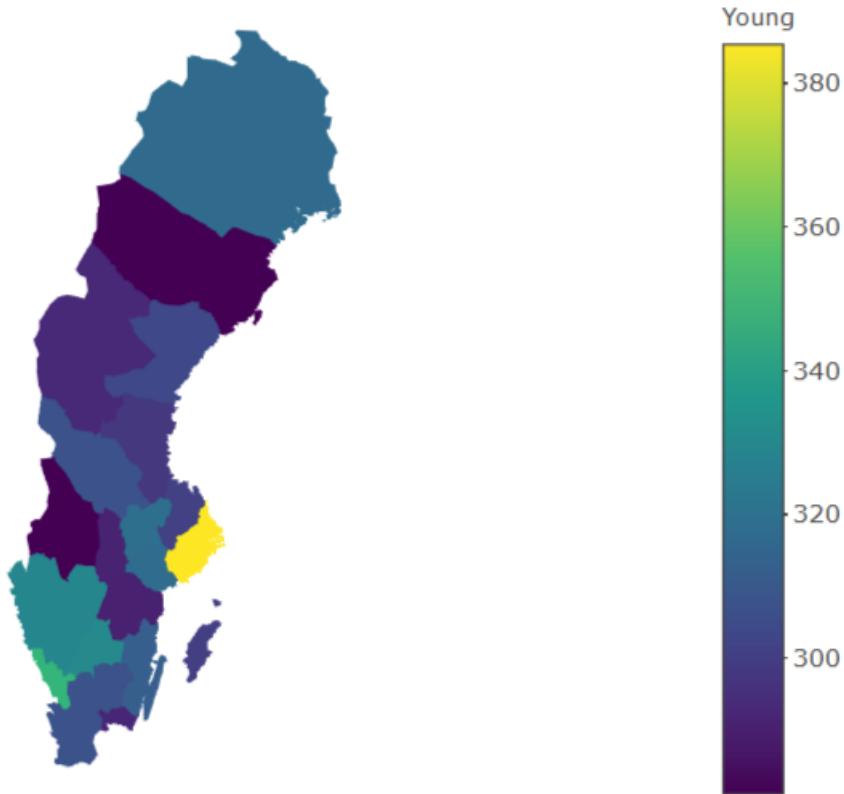
colnames(swe_data_processed) = c("region", "Young", "Adult", "Senior")
swe_data_processed$region = gsub(" county", "", swe_data_processed$region)
swe_data_processed$region = gsub("\d{2}", "", swe_data_processed$region)
swe_data_processed$region = gsub("Örebro", "Orebro", swe_data_processed$region)
rownames(swe_data_processed) = swe_data_processed$region
rds = readRDS('gadm36_SWE_1_sf.rds')

rds$Young = swe_data_processed[rds$NAME_1, "Young"]

x <- plot_ly() %>% add_sf(data = rds, split = ~NAME_1,
                           color = ~Young, showlegend = F, alpha = 1) %>%
  layout(title = "Choropleth plot of mean income of Young age group")

knitr:::include_graphics('./3.8.4.png')
```

Choropleth plot of mean income of Young age group



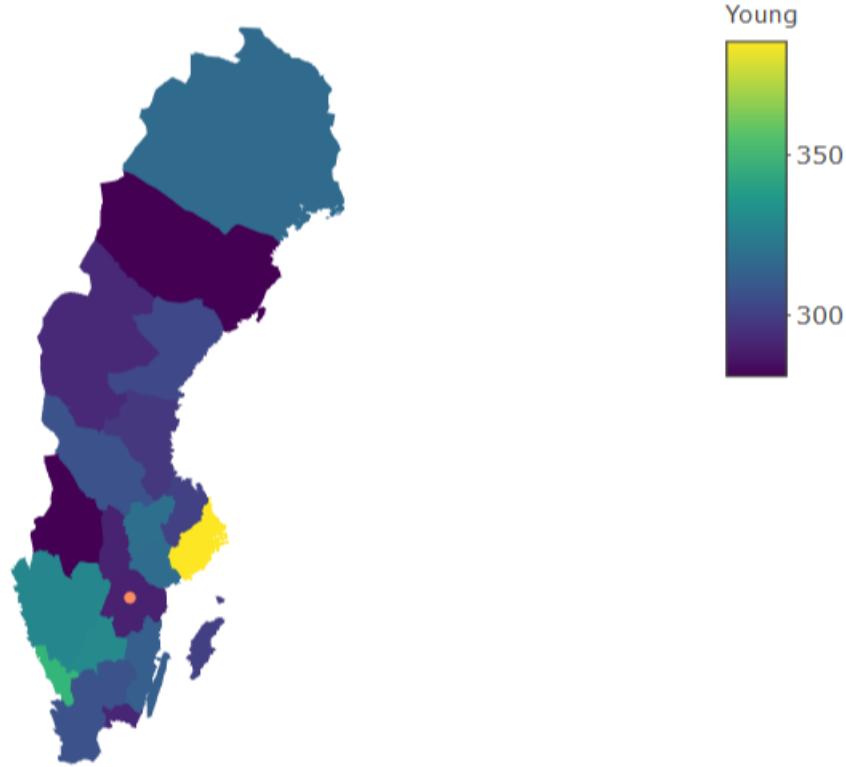
Choropleth plot with a custom marker

```
rds$Young = swe_data_processed[rds$NAME_1, "Young"]

x <- plot_ly() %>% add_sf(data = rds, split = ~NAME_1, color = ~Young,
                           showlegend = F, alpha = 1) %>%
  add_markers(x = 15.621373, y = 58.410809, color = "red",
              hoverinfo = "text", text = "We are here!") %>%
  layout(title = "Choropleth plot of mean income of Young age group")

knitr::include_graphics('./3.8.5.png')
```

Choropleth plot of mean income of Young age group



Violin Plots

Violin Plot Simple 2 variable (one categorical and one numeric)

```
swe_data_processed = data.frame(region = unique(swe_data$region))

swe_data_split = split(swe_data, swe_data$age)
for (i in seq_along(swe_data_split)) {
  swe_data_processed[[names(swe_data_split)[i]]] = merge(swe_data_split[[i]],
                                                       swe_data_processed$region,
                                                       by.x = 'region',
                                                       by.y = 1, all = T)$X2016
}

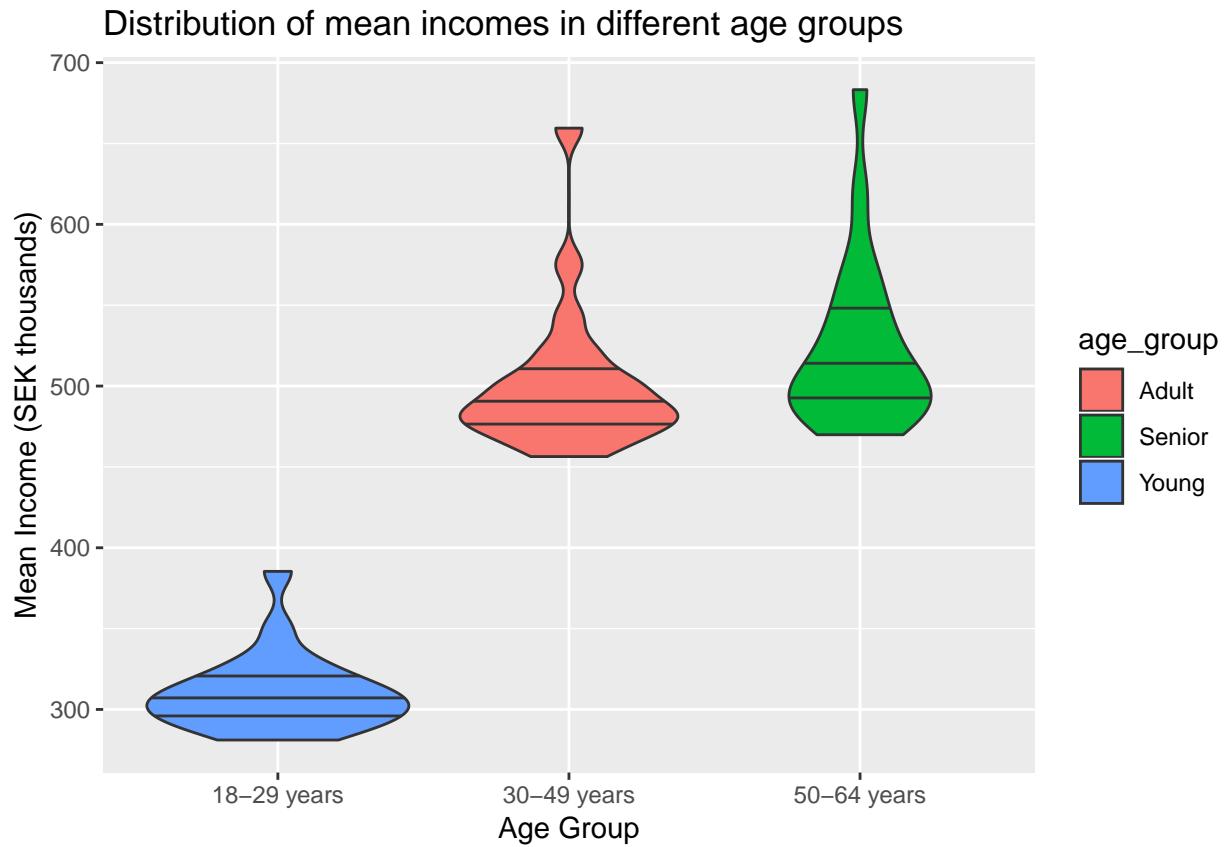
colnames(swe_data_processed) = c("region", "Young", "Adult", "Senior")
swe_data_processed$region = gsub(" county", "", swe_data_processed$region)
swe_data_processed$region = gsub("\d{2} ", "", swe_data_processed$region)
swe_data_processed$region = gsub("Örebro", "Orebro", swe_data_processed$region)
rownames(swe_data_processed) = swe_data_processed$region

swe_data$age_group <- ifelse(swe_data$age == "18-29 years", "Young", ifelse(swe_data$age == "30-49 years", "Adult", "Senior"))
```

```

ggplot(swe_data) +
  geom_violin(aes(x = age, y = X2016, fill = age_group),
              draw_quantiles = c(0.25, 0.5, 0.75)) +
  scale_x_discrete(labels=c("18_29" = "Young",
                            "30_49" = "Adult",
                            "50_69" = "Senior"),
                    name = "Age Group") +
  ylab("Mean Income (SEK thousands)") +
  ggtitle("Distribution of mean incomes in different age groups")

```



Violin using plotly

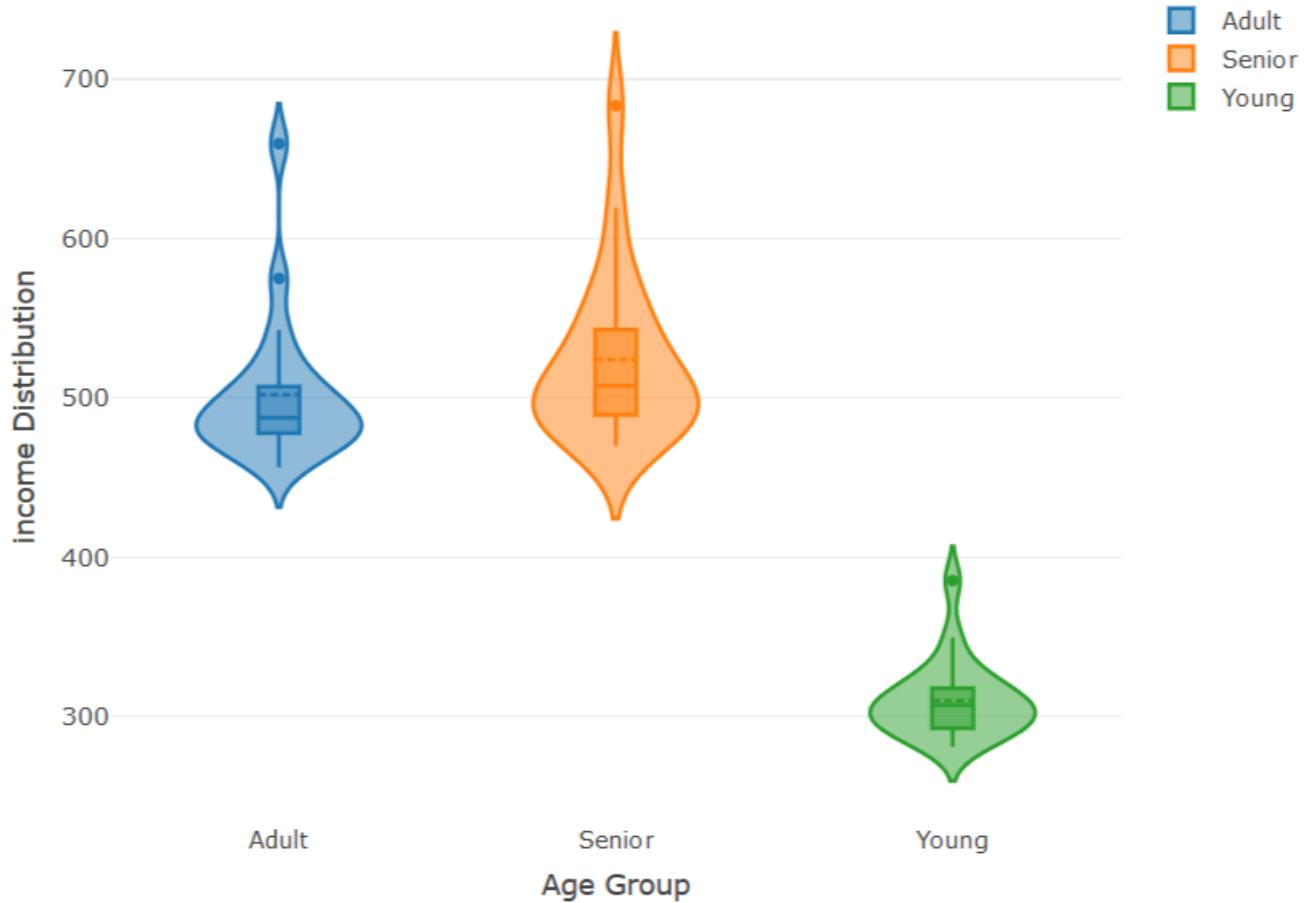
```

income_data_recasted <- dcast(income_data, region~age, value.var = "X2016")
setnames(income_data_recasted, old=c("18-29 years", "30-49 years", "50-64 years" ), new=c("Young", "Adult", "Senior"))

income_data$age_group <- ifelse(income_data$age == "18-29 years", "Young",
                                 ifelse(income_data$age == "30-49 years", "Adult", "Senior"))

x <- income_data %>% plot_ly(x = ~age_group ,y = ~X2016, type = 'violin', split = ~age_group, box = list(
  color = "#800000"))
  
```

Income Distribution vs. Age Group in Sweden 2016



3D surface plots

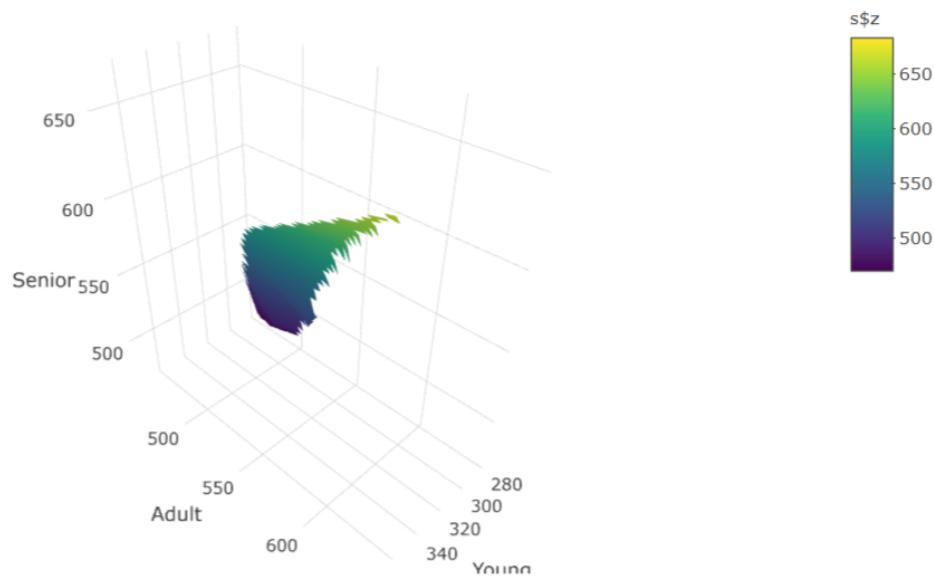
3D surface plots using plotly

```
s = interp(swe_data_processed$Young, swe_data_processed$Adult,
           swe_data_processed$Senior, duplicate = "mean")

x <- plot_ly(x=~s$x, y=~s$y, z=~s$z, type="surface") %>% layout(
  scene=list(
    xaxis = list(title = "Young"),
    yaxis = list(title = "Adult"),
    zaxis = list(title = "Senior")),
  title = "3D surface plot of income distribution")

knitr::include_graphics('./3.10.1.png')
```

3D surface plot of income distribution



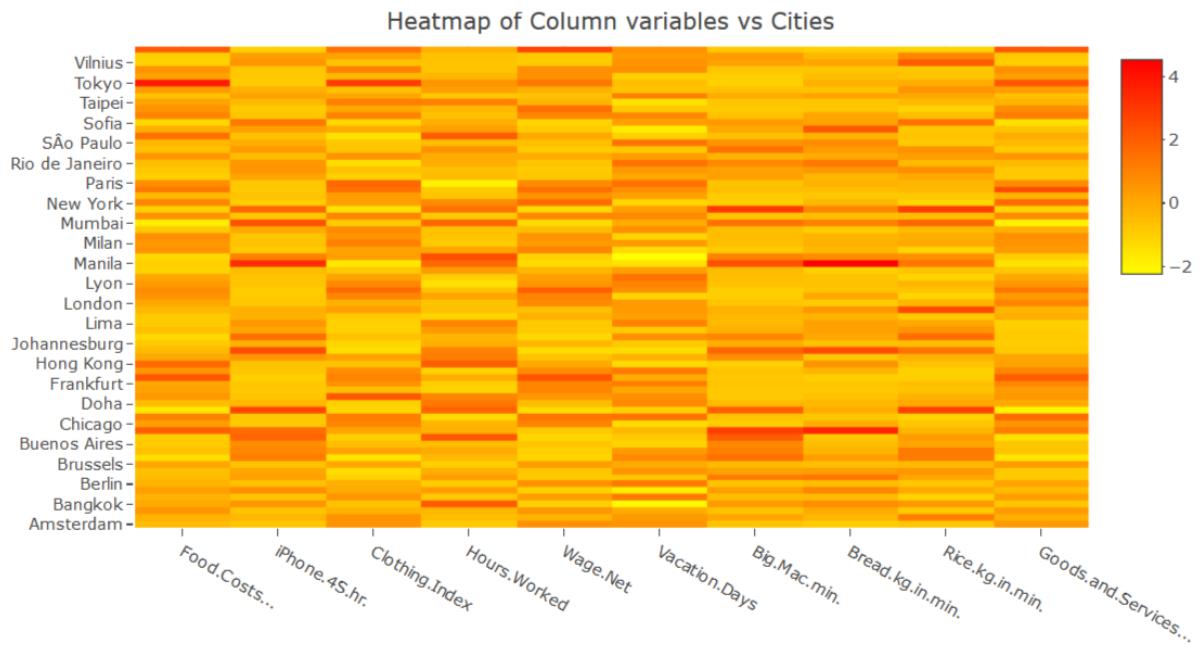
Heat Map

Heat Map without reordering

```
ubs_scaled = scale(ubs_data)

x <- plot_ly(x = colnames(ubs_scaled), y = rownames(ubs_scaled),
              z = ubs_scaled, type = "heatmap",
              colors = colorRamp(c("yellow", "red")))) %>%
  layout(title = "Heatmap of Column variables vs Cities")

knitr::include_graphics('./3.11.1.png')
```



Heat Map with ordering using Hiearchical Clustering (HC) using Euclidean distance

```

ubs_scaled = scale(ubs_data)
row_dist_euc = dist(ubs_scaled, method = "euclidean")
col_dist_euc = dist(t(ubs_scaled), method = "euclidean")

seriate_row_euc = seriate(row_dist_euc, "HC")
seriate_col_euc = seriate(col_dist_euc, "HC")
ord_row_euc = get_order(seriate_row_euc)
ord_col_euc = get_order(seriate_col_euc)

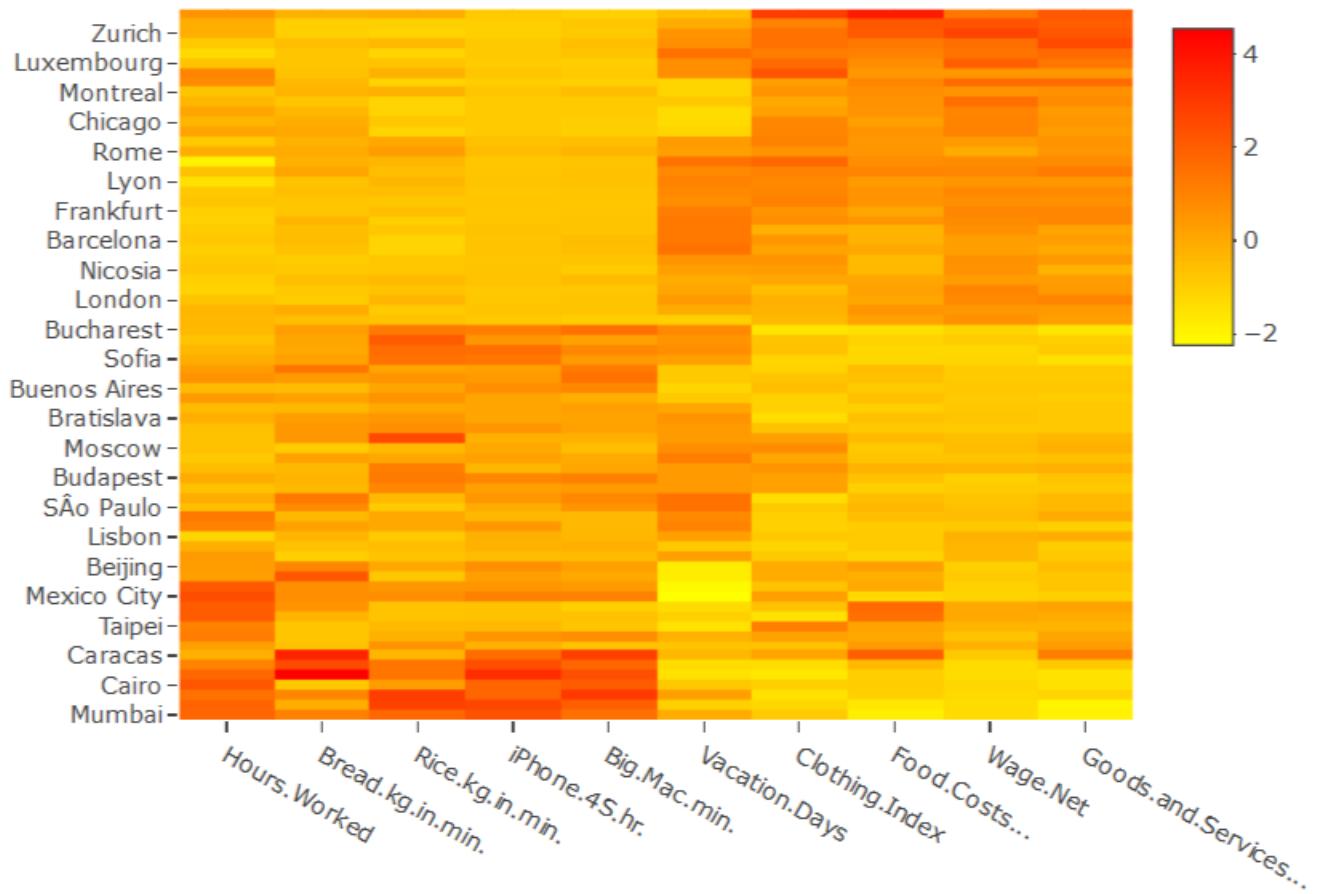
ubs_reordered_euc = ubs_scaled[rev(ord_row_euc),ord_col_euc]

x <- plot_ly(x = colnames(ubs_reordered_euc),
              y = rownames(ubs_reordered_euc),
              z = ubs_reordered_euc, type = "heatmap",
              colors = colorRamp(c("yellow", "red")))) %>%
  layout(title = "Heatmap of Column variables vs Cities")

knitr:::include_graphics('./3.11.2.png')

```

Heatmap of Column variables vs Cities



Heat Map with ordering using Corellation using Euclidean distance

```

row_dist_cor = as.dist(1 - cor(t(ubs_scaled)))
col_dist_cor = as.dist(1 - cor(ubs_scaled))

seriate_row_cor = seriate(row_dist_cor, method = "HC")
seriate_col_cor = seriate(col_dist_cor, method = "HC")
ord_row_cor = get_order(seriate_row_cor)
ord_col_cor = get_order(seriate_col_cor)

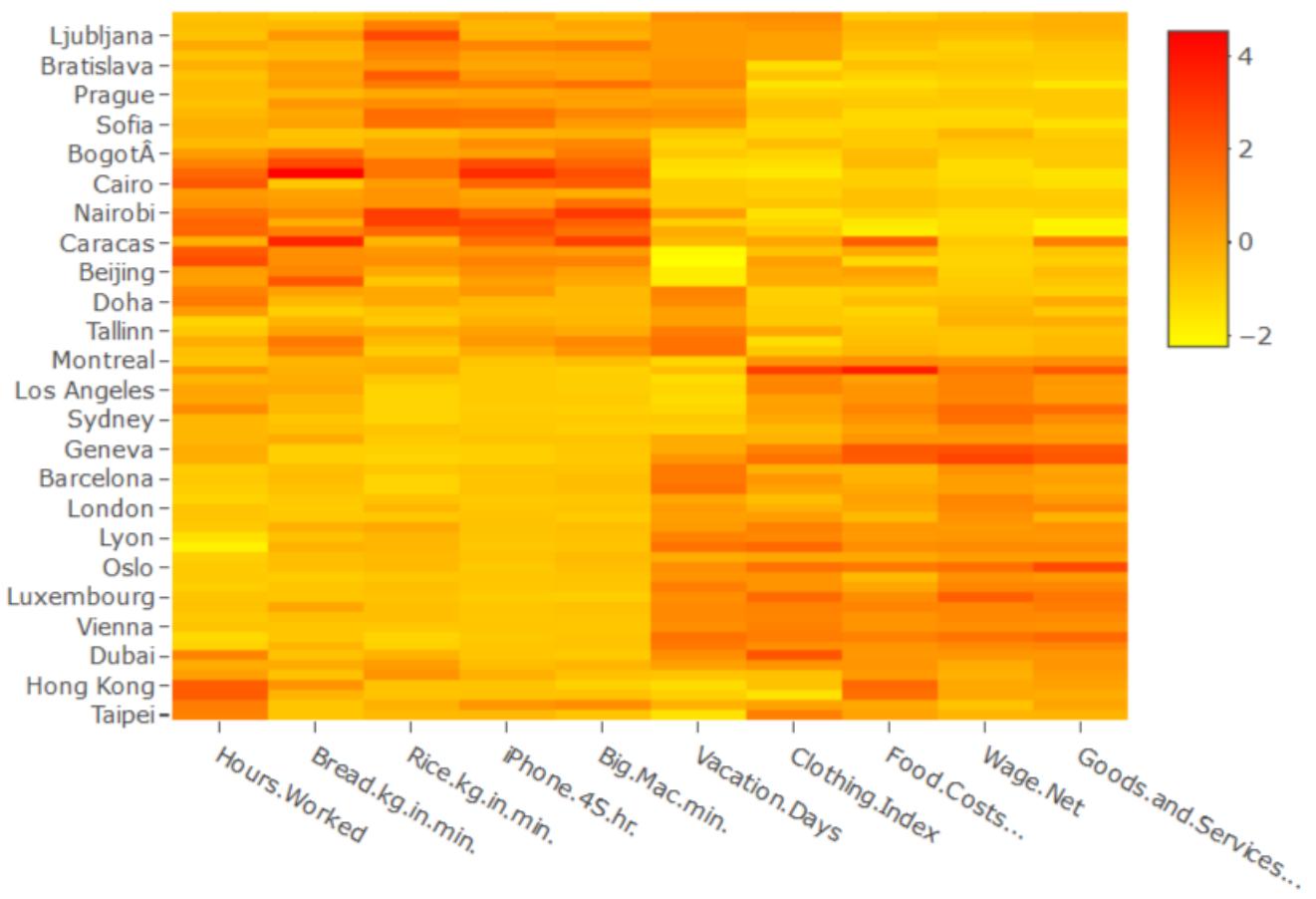
ubs_reordered_cor = ubs_scaled[rev(ord_row_cor), ord_col_cor]

x <- plot_ly(x = colnames(ubs_reordered_cor),
              y = rownames(ubs_reordered_cor),
              z = ubs_reordered_cor, type = "heatmap",
              colors = colorRamp(c("yellow", "red")))) %>%
  layout(title = "Heatmap of Column variables vs Cities")

knitr:::include_graphics('./3.11.3.png')

```

Heatmap of Column variables vs Cities



Heat Map with ordering using Hamiltonian Path Length using Traveling Salesman Problem (TSP)

```

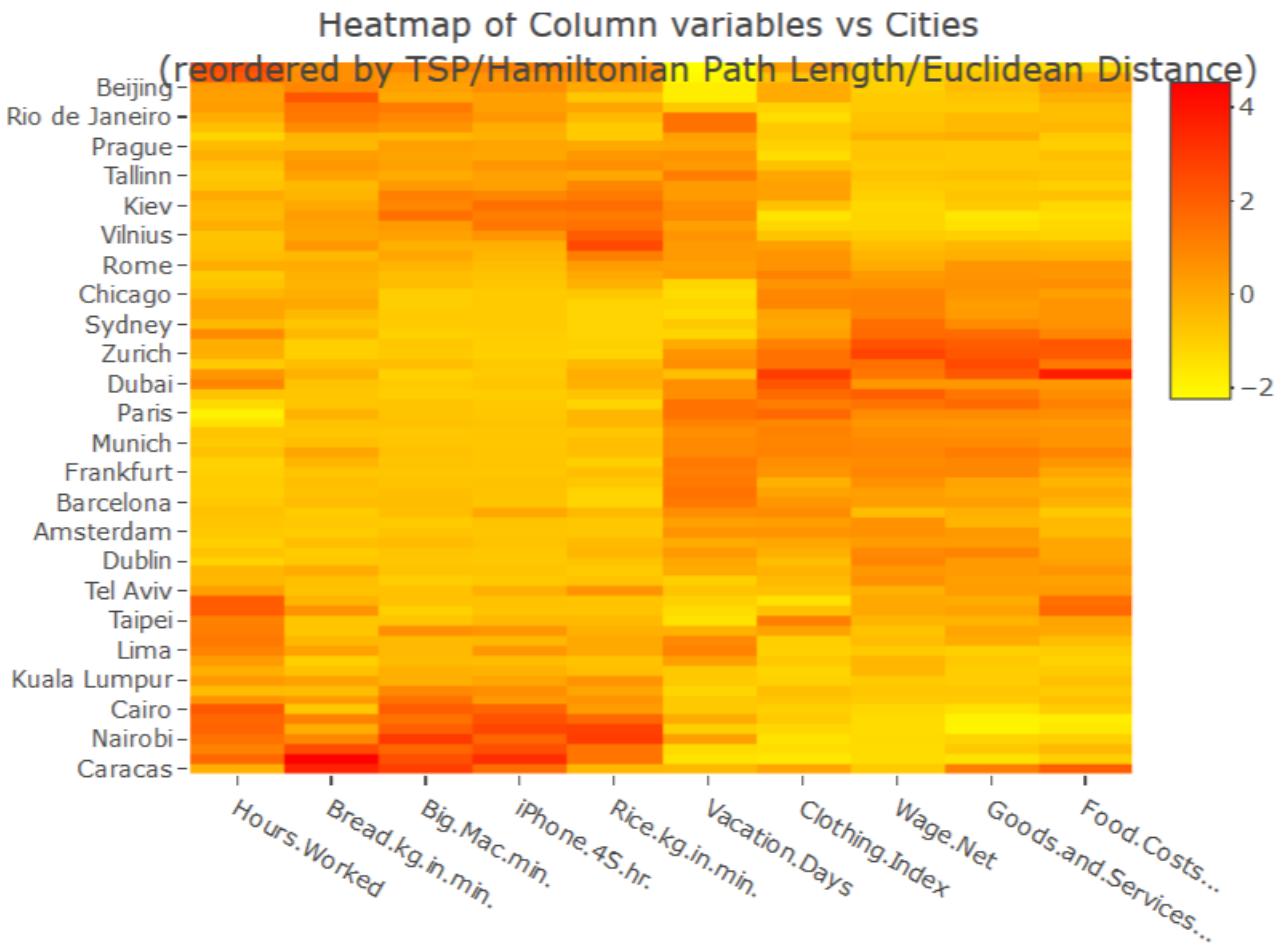
seriate_row_tsp = seriate(row_dist_euc, "TSP")
seriate_col_tsp = seriate(col_dist_euc, "TSP")
ord_row_tsp = get_order(seriate_row_tsp)
ord_col_tsp = get_order(seriate_col_tsp)

ubs_reordered_tsp = ubs_scaled[rev(ord_row_tsp), ord_col_tsp]

x <- plot_ly(x = colnames(ubs_reordered_tsp),
              y = rownames(ubs_reordered_tsp),
              z = ubs_reordered_tsp, type="heatmap",
              colors = colorRamp(c("yellow", "red")) ) %>%
  layout(title = "Heatmap of Column variables vs Cities
         (reordered by TSP/Hamiltonian Path Length/Euclidean Distance)")

knitr:::include_graphics('./3.11.4.png')

```



HeatMap QC and cluster information

```
library(heatmaply)

## Loading required package: viridis
## Loading required package: viridisLite
##
## Attaching package: 'viridis'
## The following object is masked from 'package:scales':
## 
##     viridis_pal
##
## =====
## Welcome to heatmaply version 0.15.2
## 
## Type citation('heatmaply') for how to cite the package.
## Type ?heatmaply for the main documentation.
## 
## The github page is: https://github.com/talgalili/heatmaply/
```

```

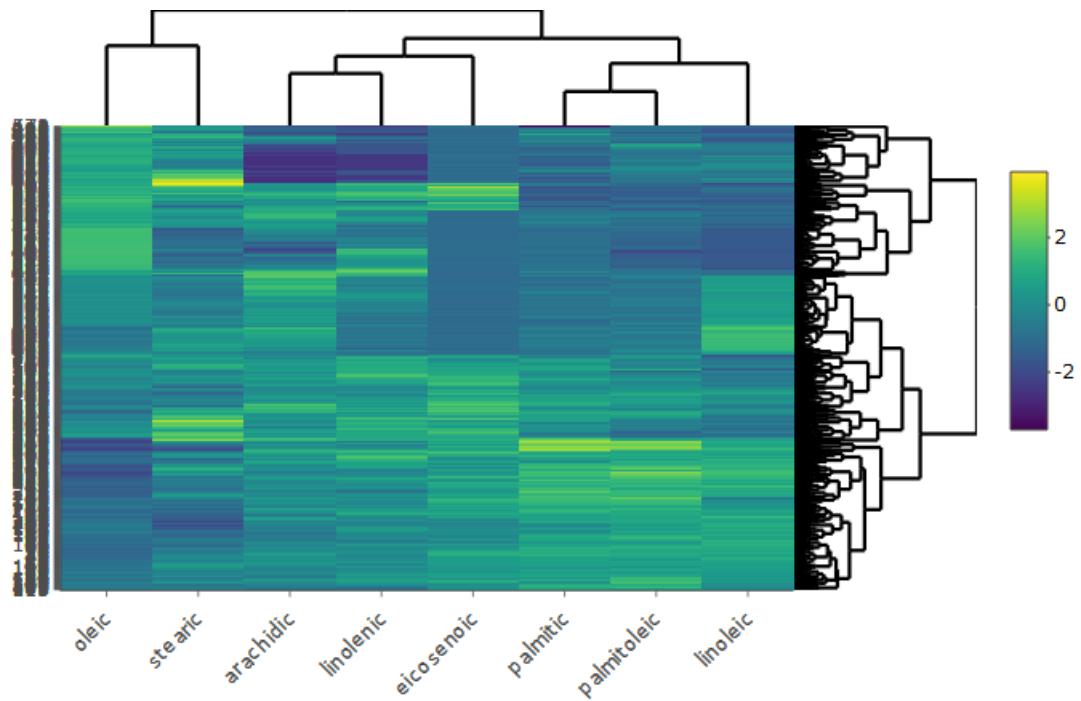
## Please submit your suggestions and bug-reports at: https://github.com/talgalili/heatmaply/issues
## Or contact: <tal.galili@gmail.com>
## =====

##
## Attaching package: 'heatmaply'

## The following object is masked from 'package:igraph':
##
##     normalize

x <- heatmaply(ubs_reordered_tsp)
knitr:::include_graphics('./heatmap_qc.png')

```



Comparison of two solvers for heatmap

```

#Hamiltonian Path Length
ord_tsp = seriate(row_dist_euc, "TSP")
ham_tsp = criterion(row_dist_euc, order = ord_tsp, "Path_length")
paste("Hamiltonian Path Length : ", ham_tsp)

## [1] "Hamiltonian Path Length : 123.487167919793"

#Gradient Measure
gm_tsp = criterion(row_dist_euc, order=ord_tsp, "Gradient_raw")
paste("Gradient Measure : ", gm_tsp)

## [1] "Gradient Measure : 40548"

#Hamiltonian Path Length
ord_hc = seriate(row_dist_euc, "HC")
ham_hc = criterion(row_dist_euc, order = ord_hc, "Path_length")

```

```

paste("Hamiltonian Path Length : ", ham_hc)

## [1] "Hamiltonian Path Length : 144.492476381981"

#Gradient Measure
gm_hc = criterion(row_dist_euc, order = ord_hc, "Gradient_raw")
paste("Gradient Measure : ", gm_hc)

## [1] "Gradient Measure : 58432"

```

Heat Map using Adjacency

```

colnames(links) <- c("from","to","strength")
links <- links[order(links$from, links$to),]
nodes$id <- 1:70
rownames(links) <- NULL

colnames(nodes)[2] <- "BombingGrp"

#Size of links based on "strength of links"
links$width <- links$strength*3

#Nodes colored based on Bombing Group
nodes$label <- nodes$V1
nodes$group <- nodes$BombingGrp

#Size of nodes proportional to the number of connections
graph <- graph.data.frame(links, directed = F)
strength_value <- strength(graph)
nodes$value <- strength_value[match(nodes$id, names(strength_value))]

for(i in 1:nrow(links)){
  links$from_name[i] <- nodes$V1[links$from[i]]
  links$to_name[i] <- nodes$V1[links$to[i]]
}

links <- links[,c("from_name","to_name","from","to","strength","width")]
nodes1 <- nodes
net <- graph_from_data_frame(d=links, vertices=nodes, directed=F)
ceb <- cluster_edge_betweenness(net)
nodes1$group <- ceb$membership

netm <- get.adjacency(net, attr="strength", sparse=F)
colnames(netm) <- V(net)$name
rownames(netm) <- V(net)$name

rowdist<-dist(netm)

order1<-seriate(rowdist, "HC")
ord1<-get_order(order1)

```

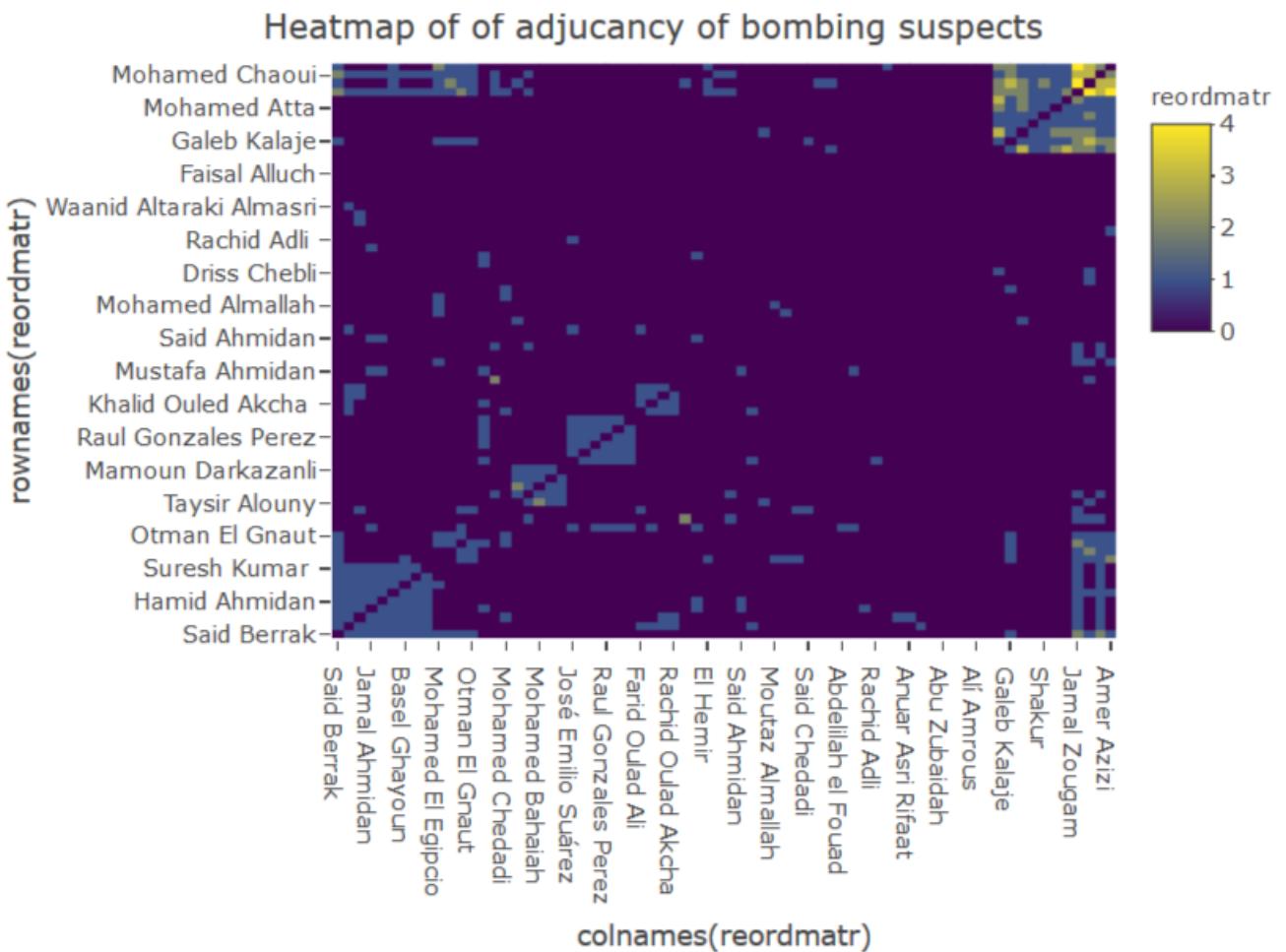
```

reordmatr<-netm[ord1,ord1]

x <- plot_ly(z=reordmatr, x=colnames(reordmatr),
              y=rownames(reordmatr), type="heatmap") %>%
layout(title = "Heatmap of of adjudicancy of bombing suspects")

knitr::include_graphics('./3.11.6.png')

```



Parallel Plot

Parallel Plot using Plotly

```

# Function to create dimension list based on given variable order
get_dimension_list <- function(df, col_order){
  dim_list = list()
  i = 1
  for (col in col_order){
    dim_list[[i]] = list(label = col, values = df[[col]])
    i = i + 1
  }
}

```

```

    return(dim_list)
}

# Create dimension list for unordered data
unord_dim_list = get_dimension_list(ubs_data, colnames(ubs_data))

# Create dimension list for reordered data
var_order = c("Cloting.Index", "Wage.Net", "Goods.and.Services...", "Food.Costs...",
             "Vacation.Days", "iPhone.4S.hr.", "Big.Mac.min.",
             "Bread.kg.in.min.", "Rice.kg.in.min.", "Hours.Worked")

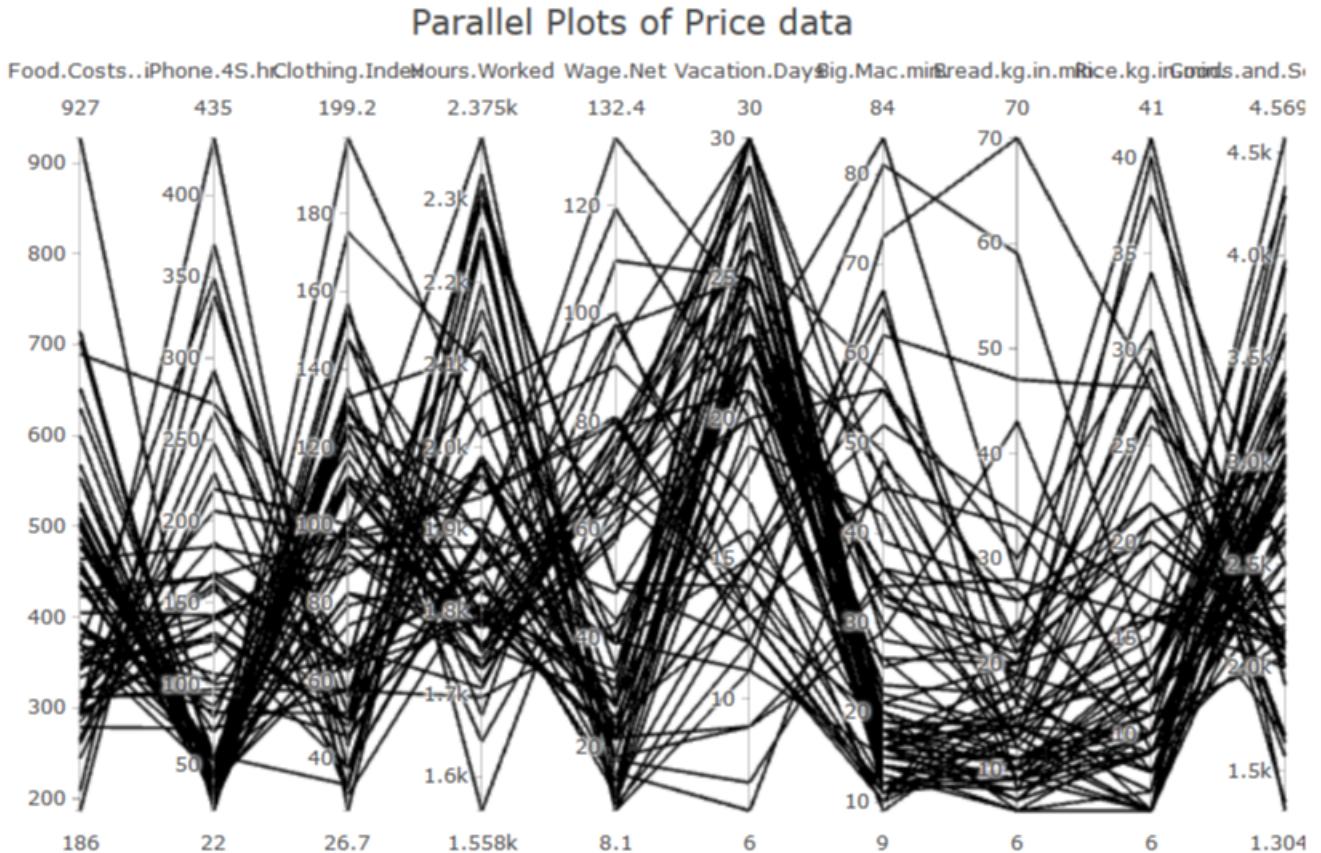
reord_dim_list = get_dimension_list(ubs_data, var_order)

# Brush different clusters with different colors
ubs_data$cluster = 1
ubs_data[ubs_data$Wage.Net < 24 , "cluster"] = 2

x <- plot_ly(data = ubs_data, type = 'parcoords',
              dimensions = unord_dim_list) %>%
  layout(title = "Parallel Plots of Price data")

knitr::include_graphics('./3.12.1.png')

```

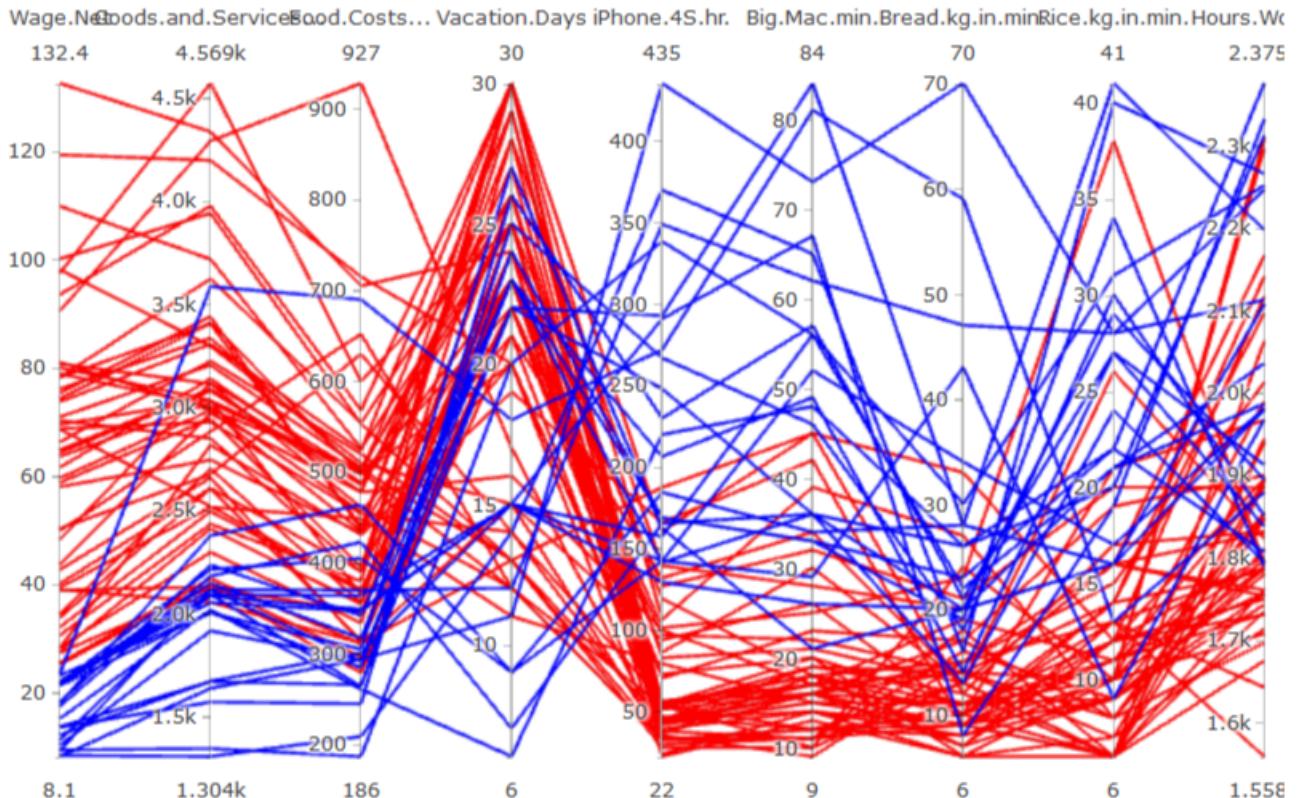


Parallel Plot using Plotly and highlight one line

```
x <- plot_ly(data = ubs_data, type = 'parcoords',
               line = list(color = ~cluster, colorscale = list(c(0,"red"), c(1,"blue"))),
               dimensions = reord_dim_list) %>%
  layout(title = "Parallel Plot with one highlight")

knitr:::include_graphics('./3.12.2.png')
```

Parallel Plot with one highlight



Radar Plots

Radar Plots using Scatterpolar

```
price_data_scale <- scale(price_data[,-1])

price_row_dist <- dist(x=price_data_scale, method = "euclidean", diag = TRUE)
price_col_dist <- dist(x=t(price_data_scale), method = "euclidean", diag = TRUE)

order1 <- seriate(price_row_dist, "OLO")
order2 <- seriate(price_col_dist, "OLO")
ord1 <- get_order(order1)
```

```
ord2 <- get_order(order2)

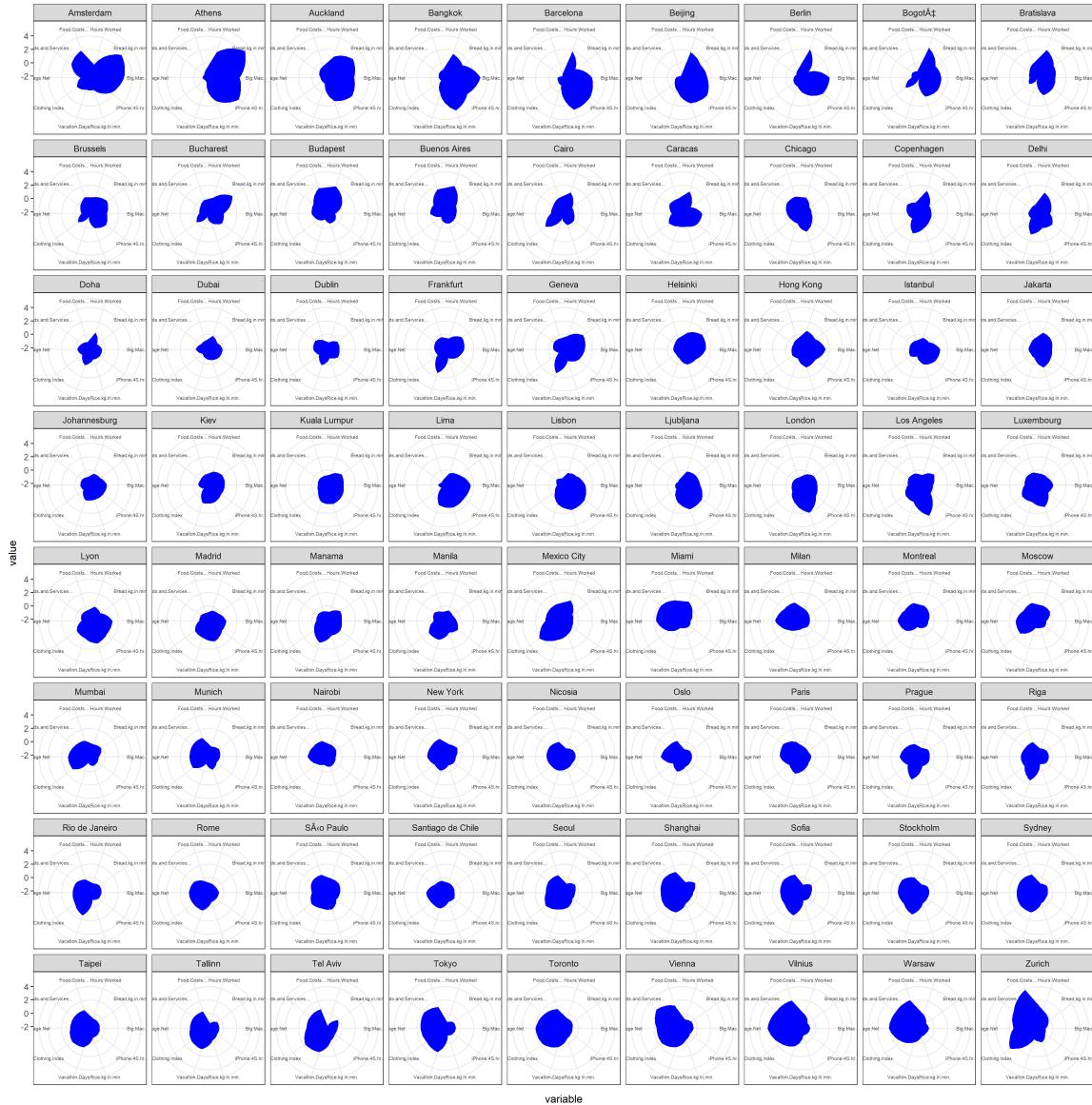
reordmatr <- price_data_scale[rev(ord1),ord2]
reordmatr <- as.data.frame(reordmatr)
reordmatr$City = price_data$City

reordmatr_transformed <- reordmatr %>% tidyverse::gather(variable, value, -City, factor_key=T) %>% arrange(City)

radar_plot <- reordmatr_transformed %>% ggplot(aes(x=variable, y=value, group=City)) + geom_polygon(fill="white", color="black", size=1)

ggsave("radar_plot.png", width = 40, height = 60, units = "cm")

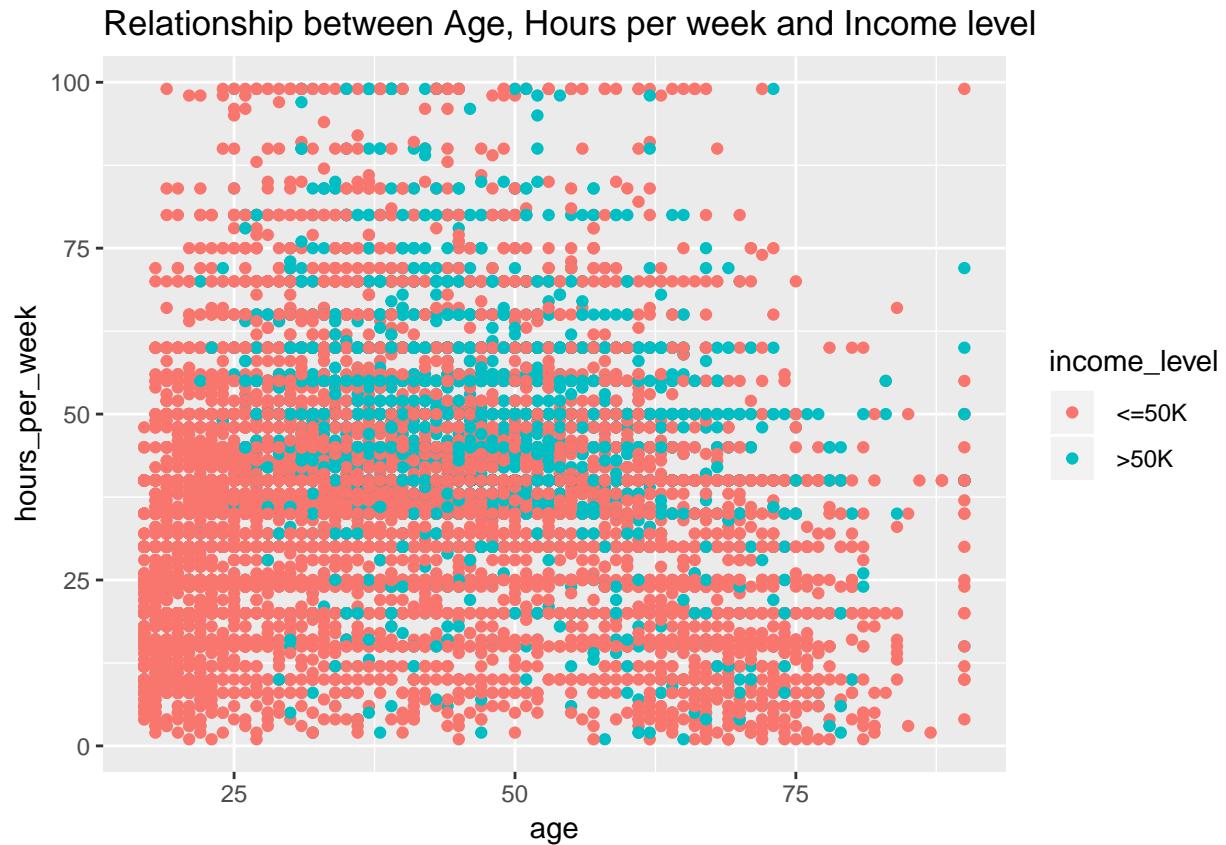
knitr::include_graphics('./radar_plot.png')
```



Trellis Plots

Trellis Plots using ggplot2

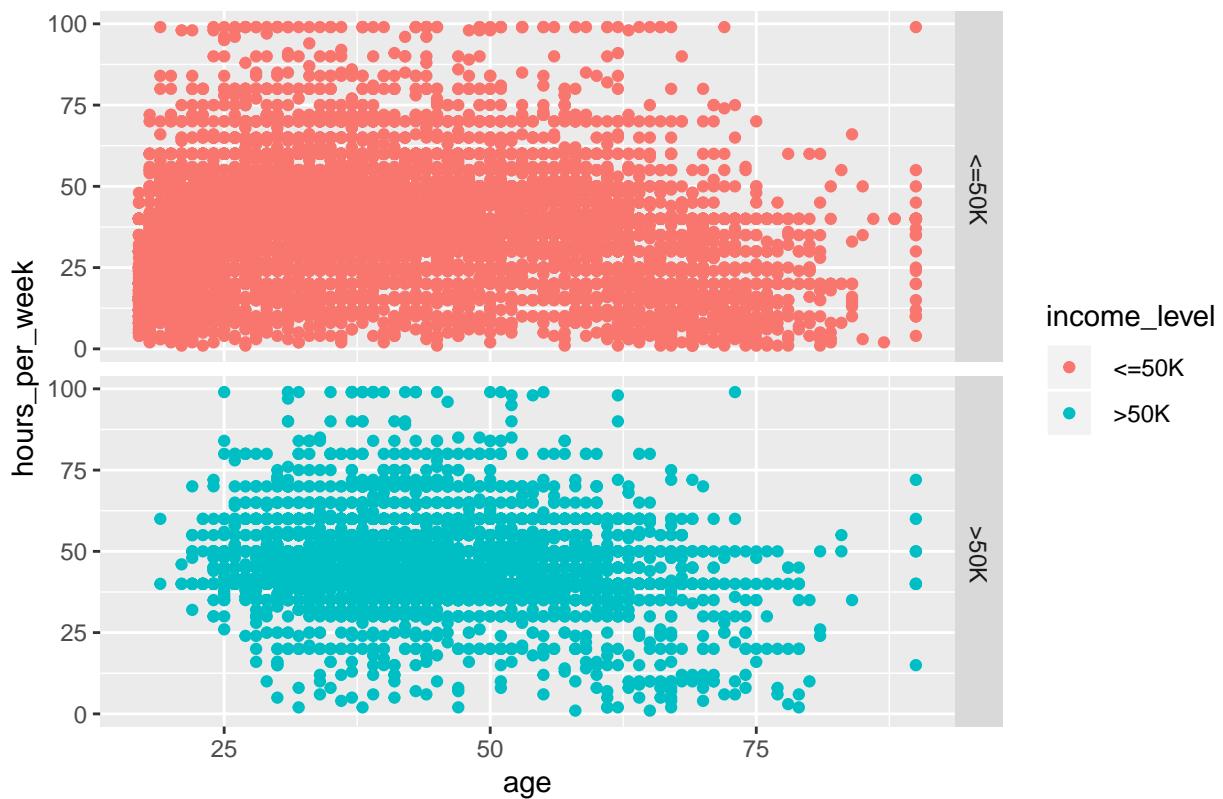
```
ggplot(adult_data) +
  geom_point(aes(x = age, y = hours_per_week, color = income_level)) +
  ggtitle("Relationship between Age, Hours per week and Income level")
```



Trellis Plots with facet/condition on a variable

```
ggplot(adult_data) +
  geom_point(aes(x = age, y = hours_per_week, color = income_level)) +
  facet_grid(income_level ~ .) +
  ggtitle("Relationship between Age, Hours per week for each Income level")
```

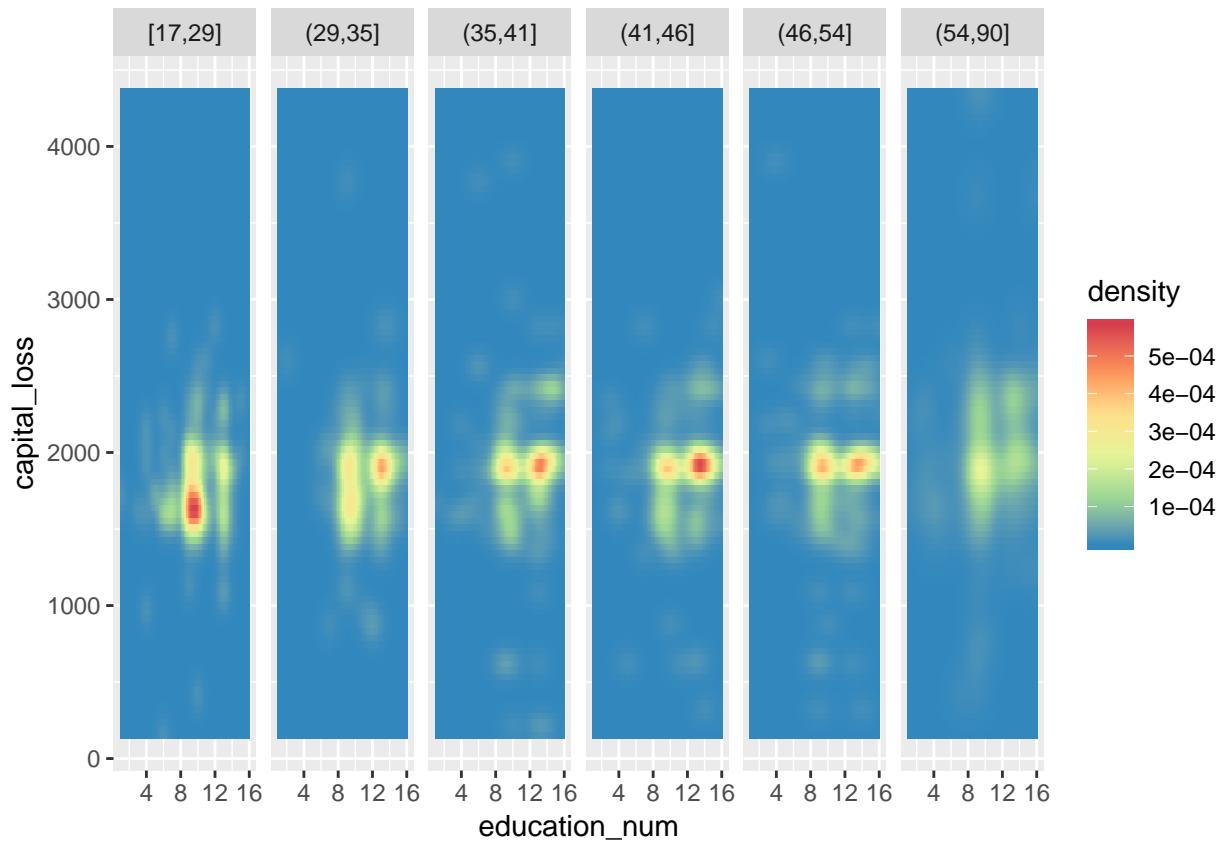
Relationship between Age, Hours per week for each Income level



Trellis Plot with raster-type-2d-density plot

```
capital_loss_data = adult_data[adult_data$capital_loss != 0,]

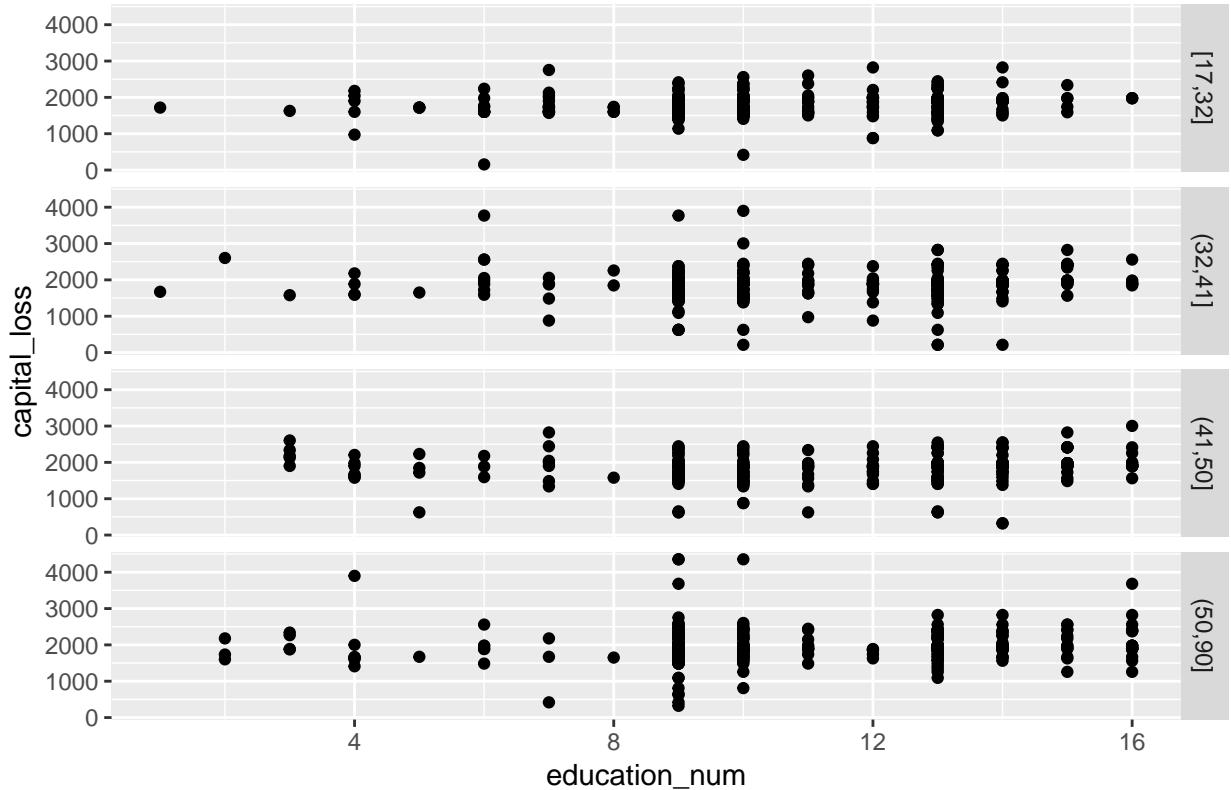
ggplot(capital_loss_data) +
  stat_density_2d(geom = "raster",
                  aes(x = education_num, y = capital_loss,
                      fill = stat(density)), contour = FALSE) +
  scale_fill_distiller(palette = "Spectral") +
  facet_grid(cols = vars(cut_number(age, 6)))
```



Trellis Plot with raster-type-2d-density plot with discretization(cut_number)

```
ggplot(capital_loss_data) +
  geom_point(aes(x = education_num, y = capital_loss)) +
  facet_grid(vars(cut_number(age, 4))) +
  ggtitle("Capital loss vs Education num for each age interval") +
  theme(plot.title = element_text(hjust = 0.5))
```

Capital loss vs Education num for each age interval



Trellis Plot with Shingles

```

age_ranges = lattice::equal.count(capital_loss_data$age, number = 4, overlap = 0.1)

age_range_matrix = matrix(unlist(levels(age_ranges)), ncol=2, byrow = T)
age_range_df = data.frame(Lower = age_range_matrix[,1],
                          Upper = age_range_matrix[,2],
                          Interval = factor(1:nrow(age_range_matrix)))

index = c()
age_interval = c()
for(i in 1:nrow(age_range_df)){
  interval_name = paste("[", age_range_df$Lower[i], ",",
                        age_range_df$Upper[i], "]", sep="")
  indices_in_interval = which(capital_loss_data$age >= age_range_df$Lower[i] &
                               capital_loss_data$age <= age_range_df$Upper[i])
  index = c(index, indices_in_interval)
  age_interval = c(age_interval, rep(interval_name, length(indices_in_interval)))
}

shingles_df = capital_loss_data[index,]
shingles_df = cbind(shingles_df, age_interval)

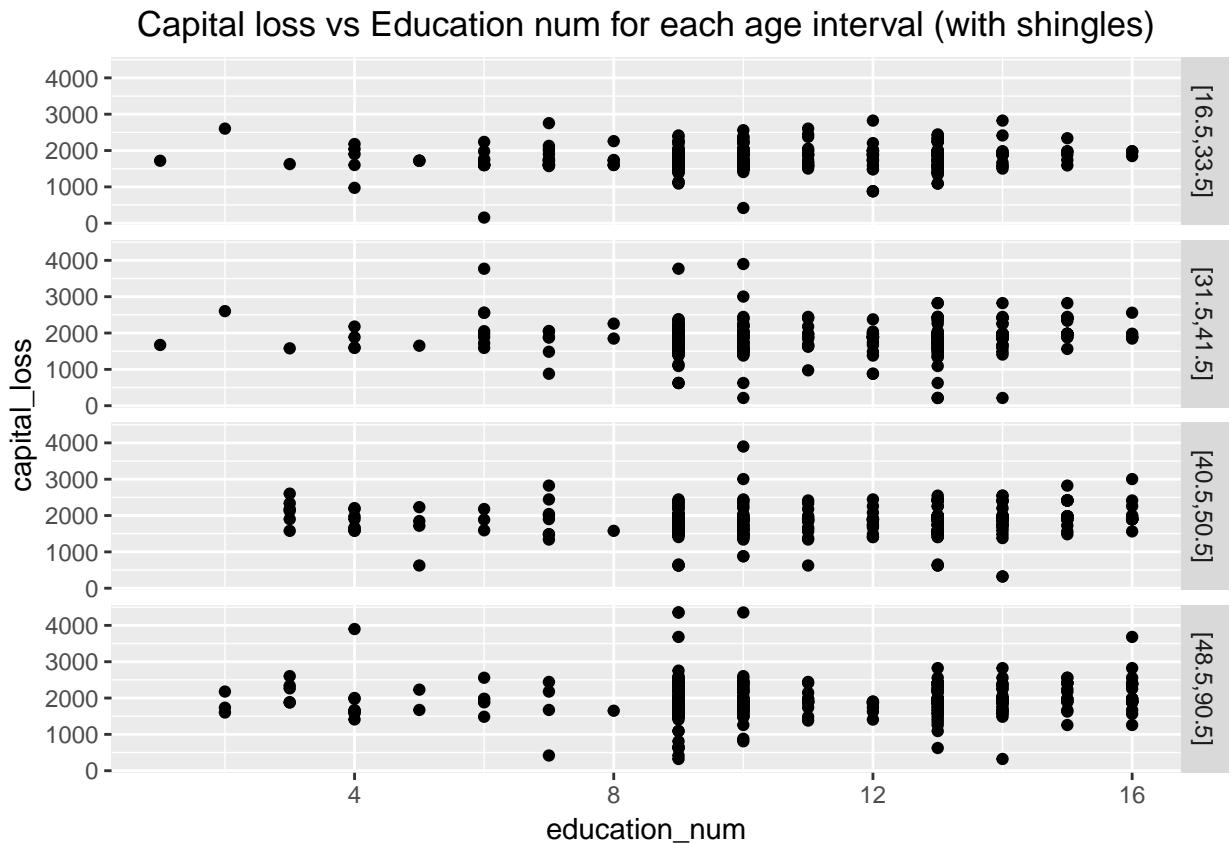
ggplot(shingles_df) +

```

```

geom_point(aes(x = education_num, y = capital_loss)) +
facet_grid(vars(age_interval)) +
ggtitle("Capital loss vs Education num for each age interval (with shingles)") +
theme(plot.title = element_text(hjust = 0.5))

```



Word Clouds

Word Clouds using tm

```

set.seed(1)
par(mfrow=c(1,2))
#Word cloud for positive reviews
positive_data$doc_id=1:nrow(positive_data)
colnames(positive_data)[1]<-"text"

#Here we interpret each line in the document as separate document
mycorpus <- Corpus(DataframeSource(positive_data)) #Creating corpus (collection of text data)
mycorpus <- tm_map(mycorpus, removePunctuation)
mycorpus <- tm_map(mycorpus, function(x) removeWords(x, stopwords("english")))
tdm <- TermDocumentMatrix(mycorpus) #Creating term-document matrix
m <- as.matrix(tdm)

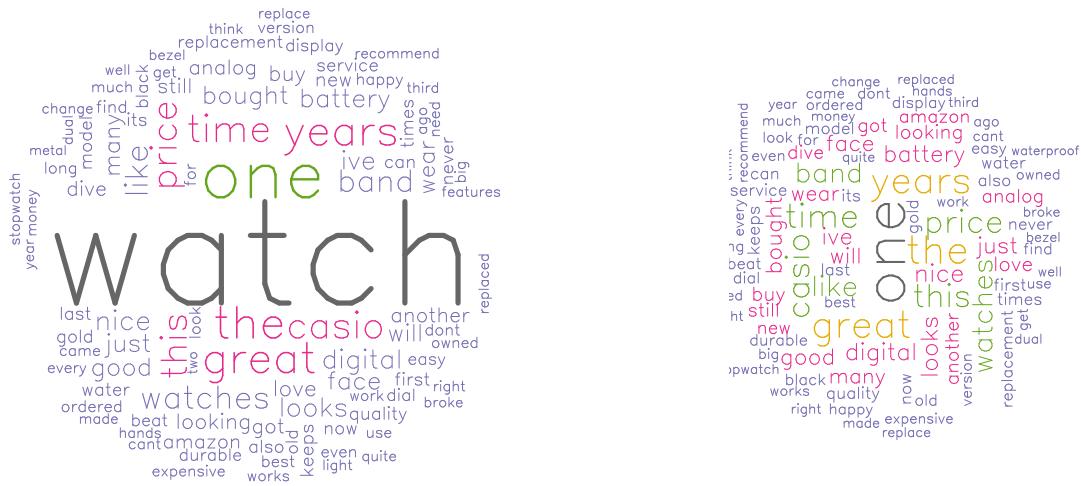
#here we merge all rows
v1 <- sort(rowSums(m),decreasing=TRUE) #Sum up the frequencies of each word

```

```
v2 <- v1[-1]

d1 <- data.frame(word = names(v1), freq=v1) #Create one column=names, second=frequencies
pal1 <- brewer.pal(8,"Dark2")
pal1 <- pal1[-(1:2)] #Create palette of colors
wordcloud(d1$word,d1$freq, scale=c(5,.3),min.freq=2,max.words=100, random.order=F, rot.per=.15, colors=pal1)

d2 <- data.frame(word = names(v2), freq=v2) #Create one column=names, second=frequencies
pal2 <- brewer.pal(8,"Dark2")
pal2 <- pal2[-(1:2)] #Create palette of colors
wordcloud(d2$word,d2$freq, scale=c(2,.3),min.freq=2,max.words=100, random.order=F, rot.per=.15, colors=pal2)
```



Linked Plots

Bar chart and Scatter Plot

Bar chart and scatter plot shared using crosstalk

```
olive_data = read.csv("olive.csv")

olive_data$Region_category <- "Sardinia Island"
olive_data[olive_data$Region == 1, c("Region_category")] <- "North"
olive_data[olive_data$Region == 2, c("Region_category")] <- "South"
```

```

ct_olive_2 = SharedData$new(olive_data)

scatter_eico_lino = plot_ly(ct_olive_2, x = ~linoleic, y = ~eicosenoic) %>%
  add_markers(color = I("black"))

bar_region <- plot_ly(ct_olive_2, x = ~Region_category) %>%
  add_histogram() %>% layout(barmode = "overlay")

plots = subplot(scatter_eico_lino, bar_region, titleX = TRUE, titleY = TRUE) %>%
  highlight(on = "plotly_select", dynamic = T, persistent = T, opacityDim = I(1)) %>%
  hide_legend() %>%
  layout(title = "Scatterplot: Eicosenoic vs Linoleic, Histogram: Region")

## Adding more colors to the selection color palette.

## We recommend setting `persistent` to `FALSE` (the default) because persistent selection mode can now
x <- bscols(widths = c(12), list(tags$h2("Analysis of variable relationships"),
                                    tags$h4("(Eicosenoic, Linoleic, Region, Stearic)"),
                                    filter_slider("stearic", "Stearic", ct_olive_2, ~stearic), plots))

knitr::include_graphics('./4.1.1.png')

```

Analysis of variable relationships

(Eicosenoic, Linoleic, Region, Stearic)

Stearic

152

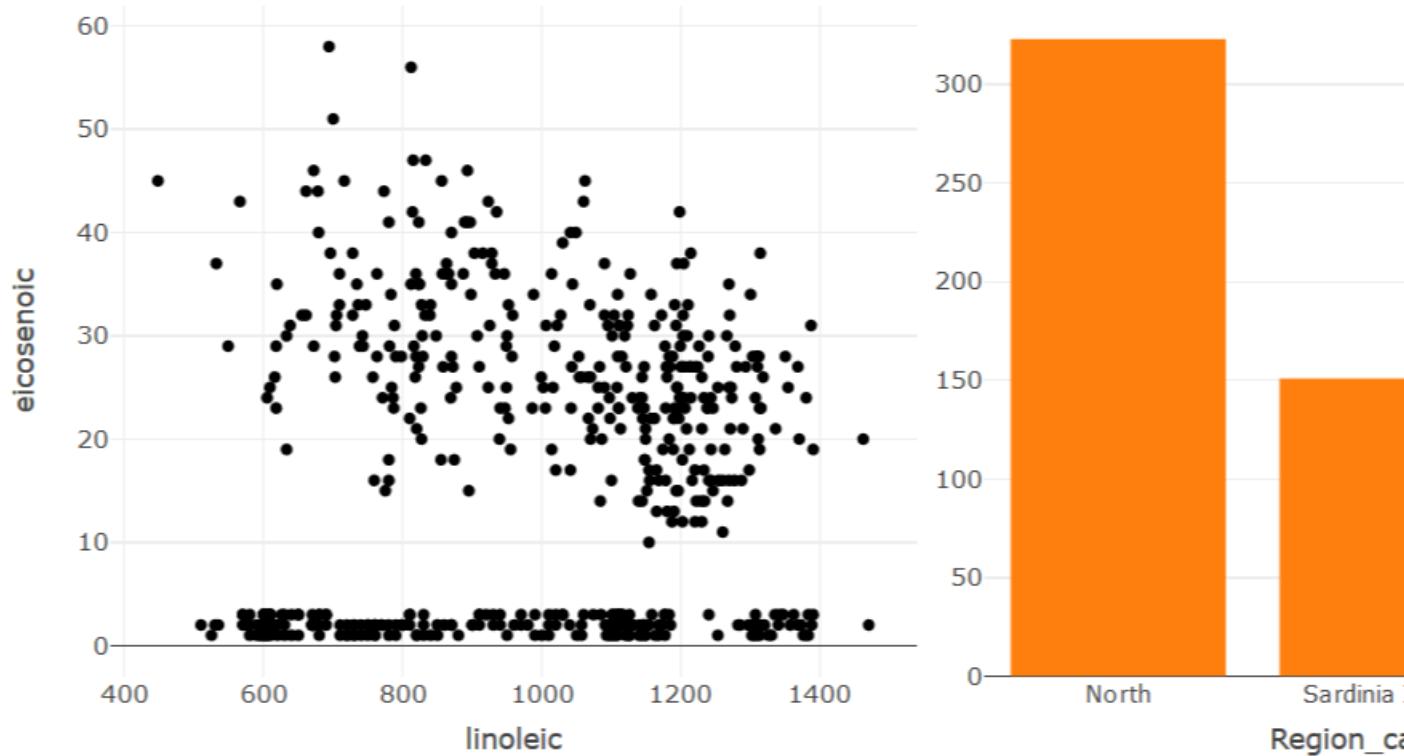


152 176 200 224 248 272 296 320

Brush color

rgba(228

Scatterplot: Eicosenoic vs Linoleic, Histogram: Region



Scatter plot shared using crosstalk

```
ct_olive_3 = SharedData$new(olive_data)

sctr_eico_lino = plot_ly(ct_olive_3, x = ~linoleic, y = ~eicosenoic) %>%
  add_markers(color = I("black"))

sctr_arac_lino = plot_ly(ct_olive_3, x = ~linolenic, y = ~arachidic) %>%
  add_markers(color = I("black"))
```

```

x <- subplot(sctr_eico_lino, sctr_arac_lino, titleX = TRUE, titleY = TRUE) %>%
  highlight(on = "plotly_select", dynamic = T, persistent = T, opacityDim = I(1)) %>%
  hide_legend() %>%
  layout(title = "Scatterplots: Eicosenoic vs Linoleic, Arachidic vs Linolenic")

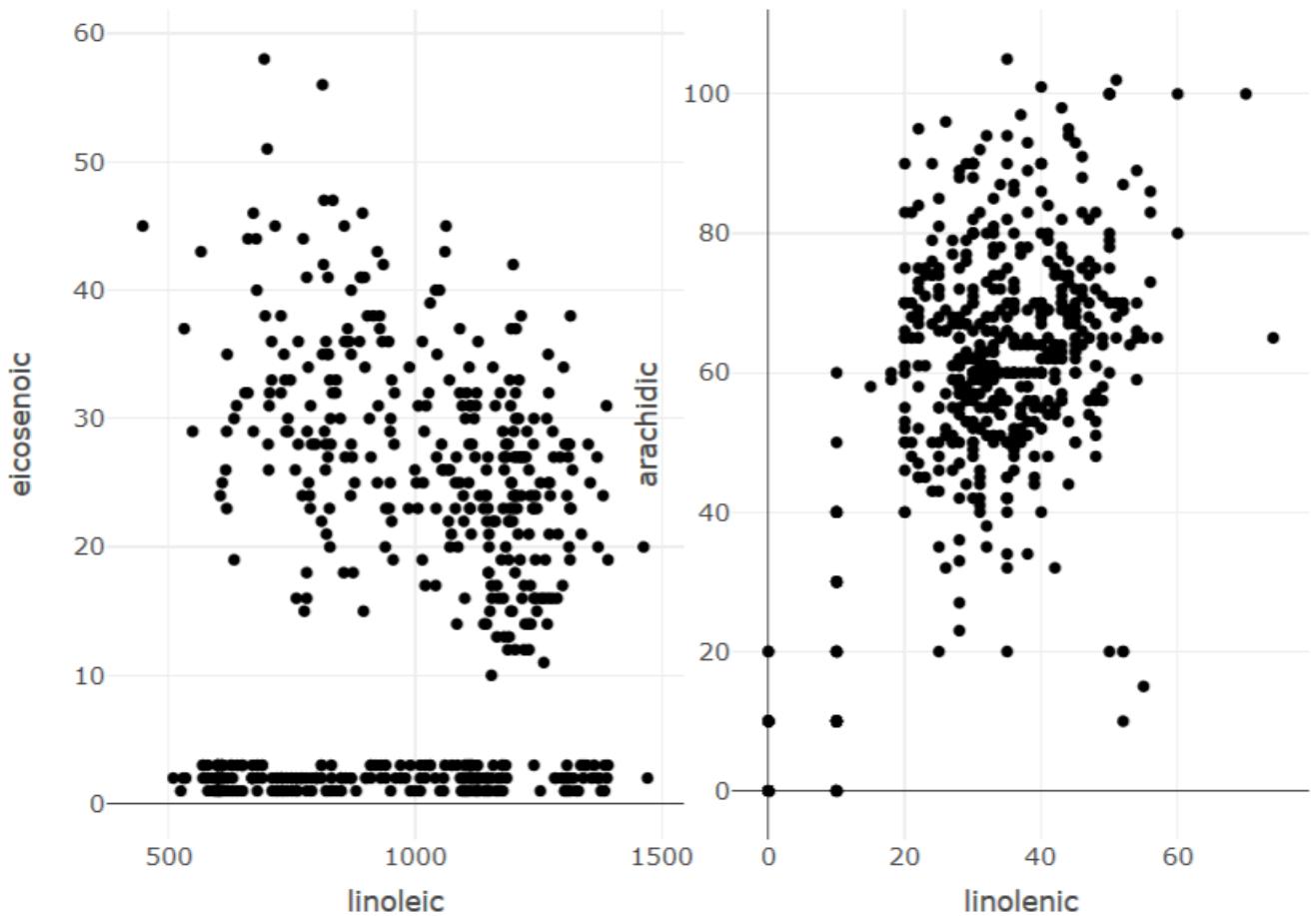
knitr:::include_graphics('./4.1.2.png')

```

Brush color

rgba(228

Scatterplots: Eicosenoic vs Linoleic, Arachidic vs Linolenic



Parallel Chord/Trellis Plot Scatter plot and Bar chart

```

var_order = c("palmitic", "palmitoleic", "linoleic", "stearic",
            "oleic", "linolenic", "arachidic", "eicosenoic")

par_plot = ggparcoord(olive_data[var_order], columns = 1:8)

plot_data = plotly_data(ggplotly(par_plot)) %>% group_by(.ID)
ct_olive_4 = SharedData$new(plot_data, ~.ID, group = "olive")
par_coord_plot = plot_ly(ct_olive_4, x = ~variable, y = ~value) %>%

```

```

add_lines(line = list(width = 0.3))%>%
add_markers(marker = list(size = 0.3),
            text = ~.ID, hoverinfo = "text") %>%
layout(title = "Parallel Coordinate Plots")

ct_olive = SharedData$new(olive_data, group = "olive")

get_buttons = function(df, axis){
  buttons = list()
  i = 1
  for(col in colnames(df)){
    buttons[[i]] = list(method = "restyle",
                        args = list(axis, list(olive_data[[col]])),
                        label = paste(axis , ": ", col, sep=""))
    i = i + 1
  }

  return(buttons)
}

buttons_x = get_buttons(olive_data[, 4:11], "x")
buttons_y = get_buttons(olive_data[, 4:11], "y")
buttons_z = get_buttons(olive_data[, 4:11], "z")

annot = list(list(text = "X", x=-0.6, y = 0.78, xref = 'paper', yref = 'paper', showarrow = FALSE),
             list(text = "Y", x=-0.6, y = 0.55, xref = 'paper', yref = 'paper', showarrow = FALSE),
             list(text = "Z", x=-0.6, y = 0.34, xref = 'paper', yref = 'paper', showarrow = FALSE))

scatter_plot_3d = plot_ly(ct_olive, x = ~palmitic, y = ~palmitoleic, z = ~stearic) %>%
  add_markers(marker = list(size=4), opacity = 0.5) %>%
  layout(scene = list(xaxis = list(title = "X"),
                      yaxis = list(title = "Y"),
                      zaxis = list(title = "Z")),
         title = "3D Scatterplot",
         # annotations = annot,
         updatemenus = list(
           list(y = 0.4, buttons = buttons_x, text = "X", active = 0),
           list(y = 0.6, buttons = buttons_y, text = "Y", active = 1),
           list(y = 0.8, buttons = buttons_z, text = "Z", active = 2)))

bar_chart = plot_ly(ct_olive, x = ~Region_category) %>%
  add_histogram() %>% layout(title = "Histogram: Region", barmode = "overlay")

x <- bscols(widths = c(12, 12, 6, 6),
tags$h2("Interactive plots to analyze relationships between variables"),
  par_coord_plot %>%
    highlight(on = "plotly_select", dynamic = T, persistent = T, opacityDim = I(1)) %>%
    hide_legend(),
  scatter_plot_3d %>%
    highlight(on = "plotly_click", dynamic = T, persistent = T) %>% hide_legend(),
  bar_chart %>%
    highlight(on = "plotly_click", dynamic = T, persistent = T) %>% hide_legend())

```

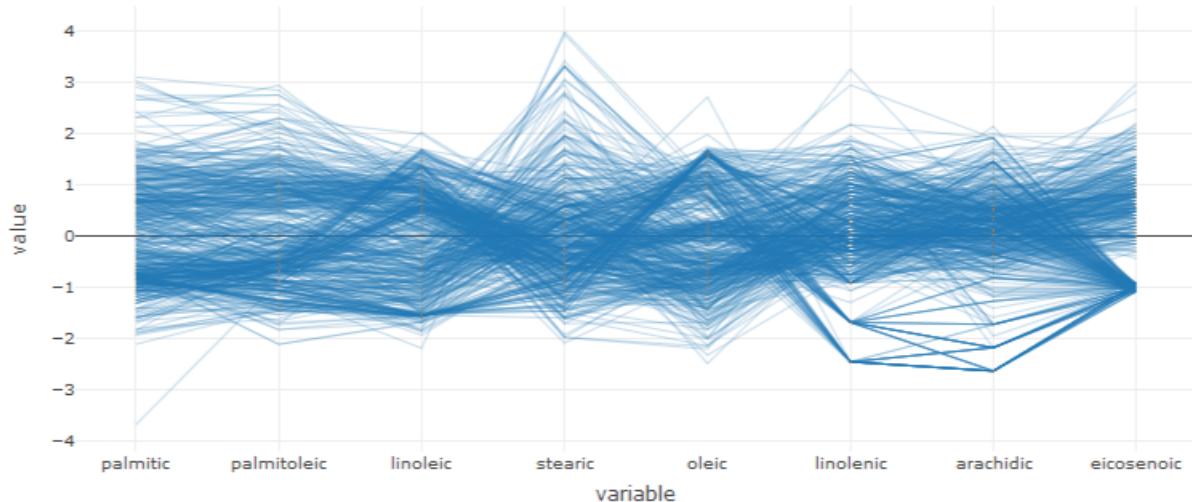
```
knitr::include_graphics('4.1.3.png')
```

Interactive plots to analyze relationships between variables

Brush color

rgba(228)

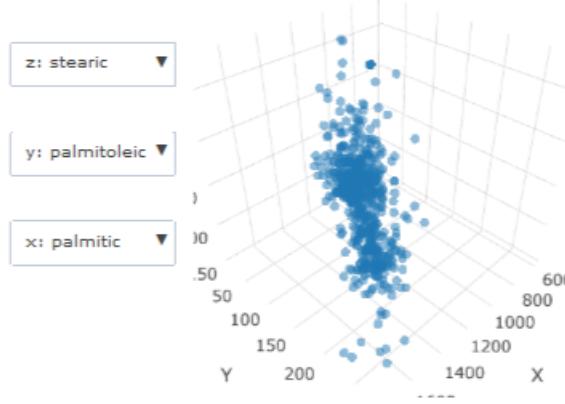
Parallel Coordinate Plots



Brush color

rgba(228)

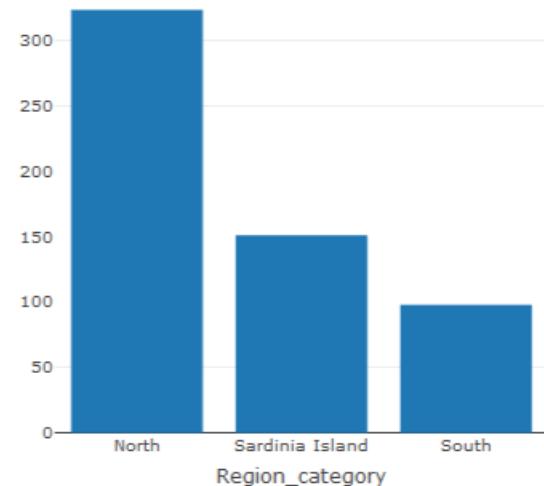
3D Scatterplot



Brush color

rgba(228)

Histogram: Region



Network Graphs

Network Graph using visNetwork

Replusion Optimzed Network Graph

```
links <- read.table("trainData.dat", as.is=T)
nodes <- as.data.frame(read.table("trainMeta.dat", as.is=T))

colnames(links) <- c("from","to","strength")
links <- links[order(links$from, links$to),]
nodes$id <- 1:70
rownames(links) <- NULL

colnames(nodes)[2] <- "BombingGrp"

#Size of links based on "strength of links"
links$width <- links$strength*3

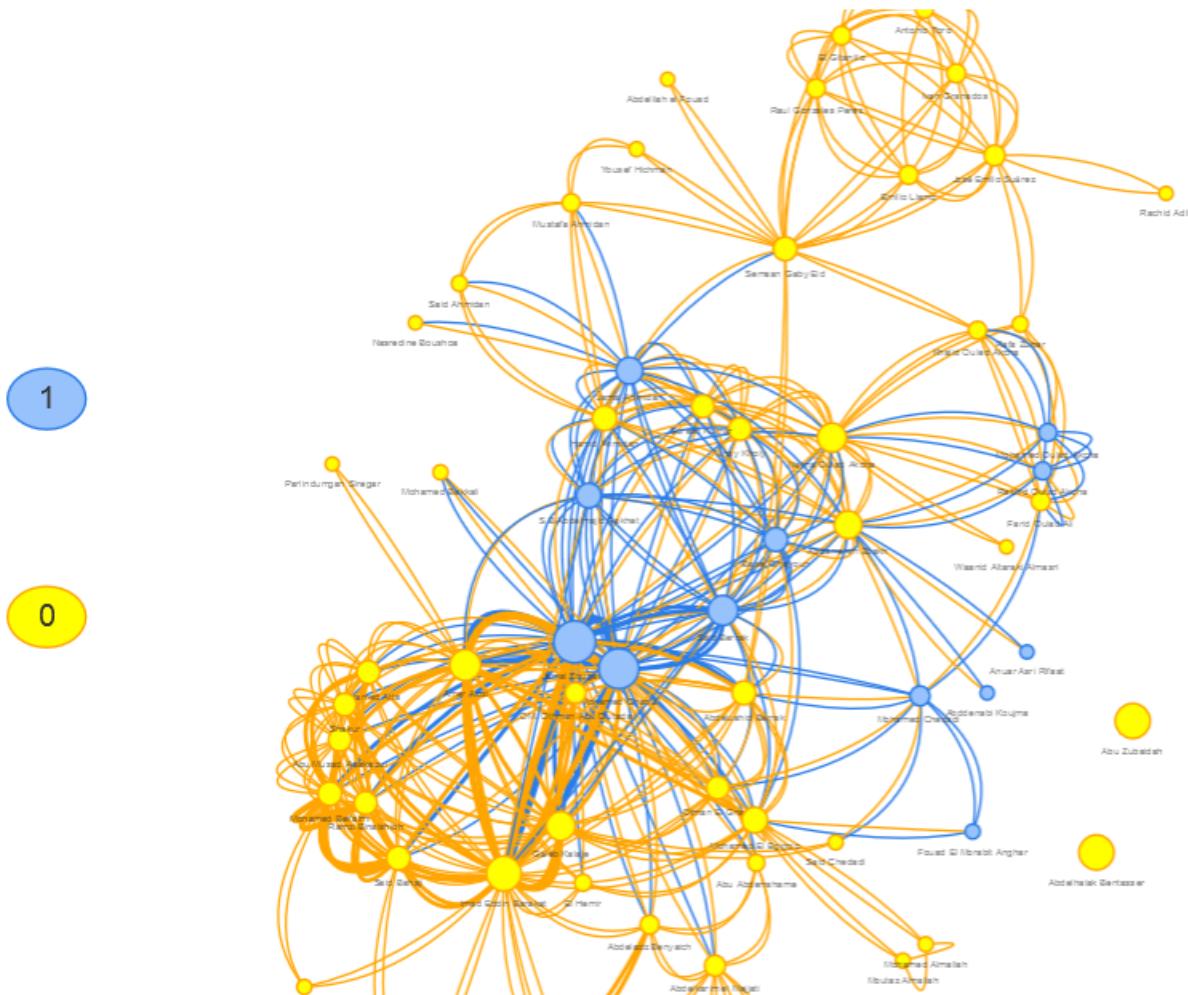
#Nodes colored based on Bombing Group
nodes$label <- nodes$V1
nodes$group <- nodes$BombingGrp

#Size of nodes proportional to the number of connections
graph <- graph.data.frame(links, directed = F)
strength_value <- strength(graph)
nodes$value <- strength_value[match(nodes$id, names(strength_value))]

x <- visNetwork(nodes, links, main = "Network of people involved in Madrid Bombing") %>%
  visLegend() %>%
  visLayout(randomSeed = 8) %>%
  visPhysics(solver="repulsion") %>%
  visOptions(highlightNearest = list(enabled = T, degree = 1, hover = T))

knitr:::include_graphics('./5.1.1.png')
```

Network of people involved in Madrid Bombing



Repulsion Optimized Network Graph with highlights

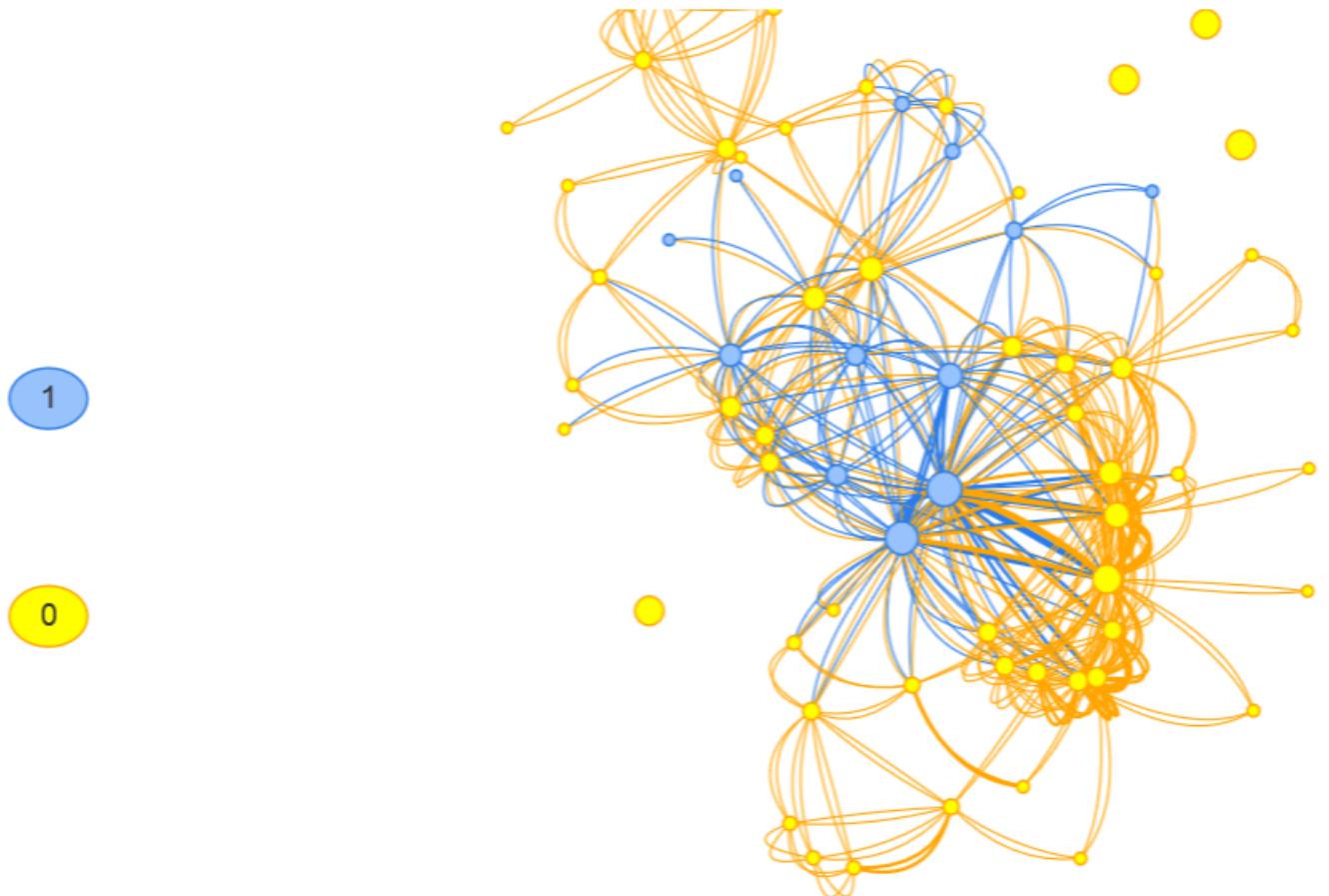
```

x <- visNetwork(nodes, links, main = "Network of people involved in Madrid Bombing") %>%
  visLegend() %>%
  visLayout(randomSeed = 12) %>%
  visPhysics(solver="repulsion") %>%
  visOptions(highlightNearest = list(enabled = T, degree = 2, hover = T))

knitr:::include_graphics('./5.1.2.png')

```

Network of people involved in Madrid Bombing



Edge Between Optimizing Network Graph

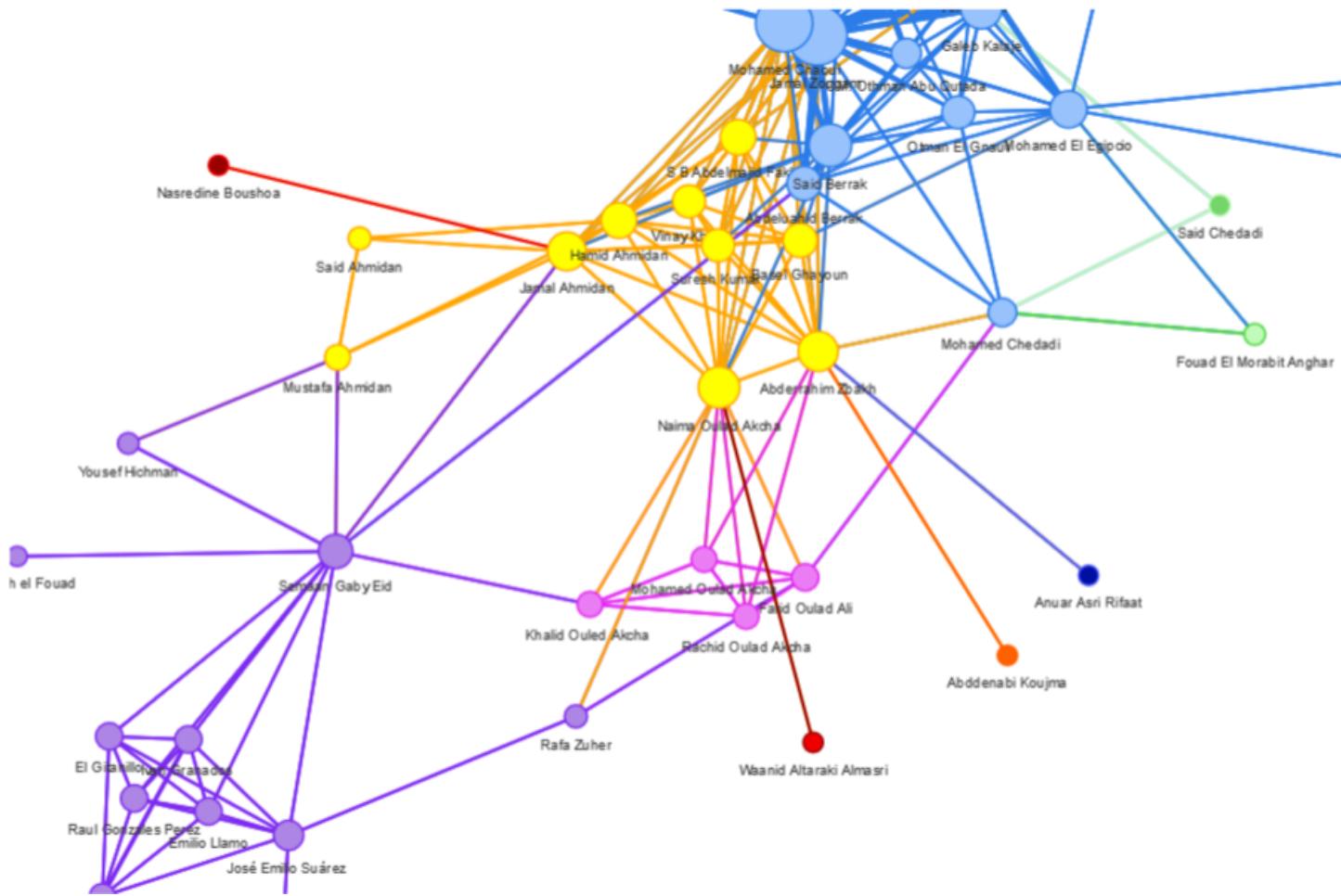
```
for(i in 1:nrow(links)){
  links$from_name[i] <- nodes$V1[links$from[i]]
  links$to_name[i] <- nodes$V1[links$to[i]]
}

links <- links[,c("from_name","to_name","from","to","strength","width")]
nodes1 <- nodes
net <- graph_from_data_frame(d=links, vertices=nodes, directed=F)
ceb <- cluster_edge_betweenness(net)
nodes1$group <- ceb$membership

x <- visNetwork(nodes1,links, main = "Network of people involved in Madrid Bombing") %>%
  visIgraphLayout() %>%
  visOptions(highlightNearest = list(enabled = T, degree = 2, hover = T))

knitr:::include_graphics('./5.1.3.png')
```

Network of people involved in Madrid Bombing



Animated Plots

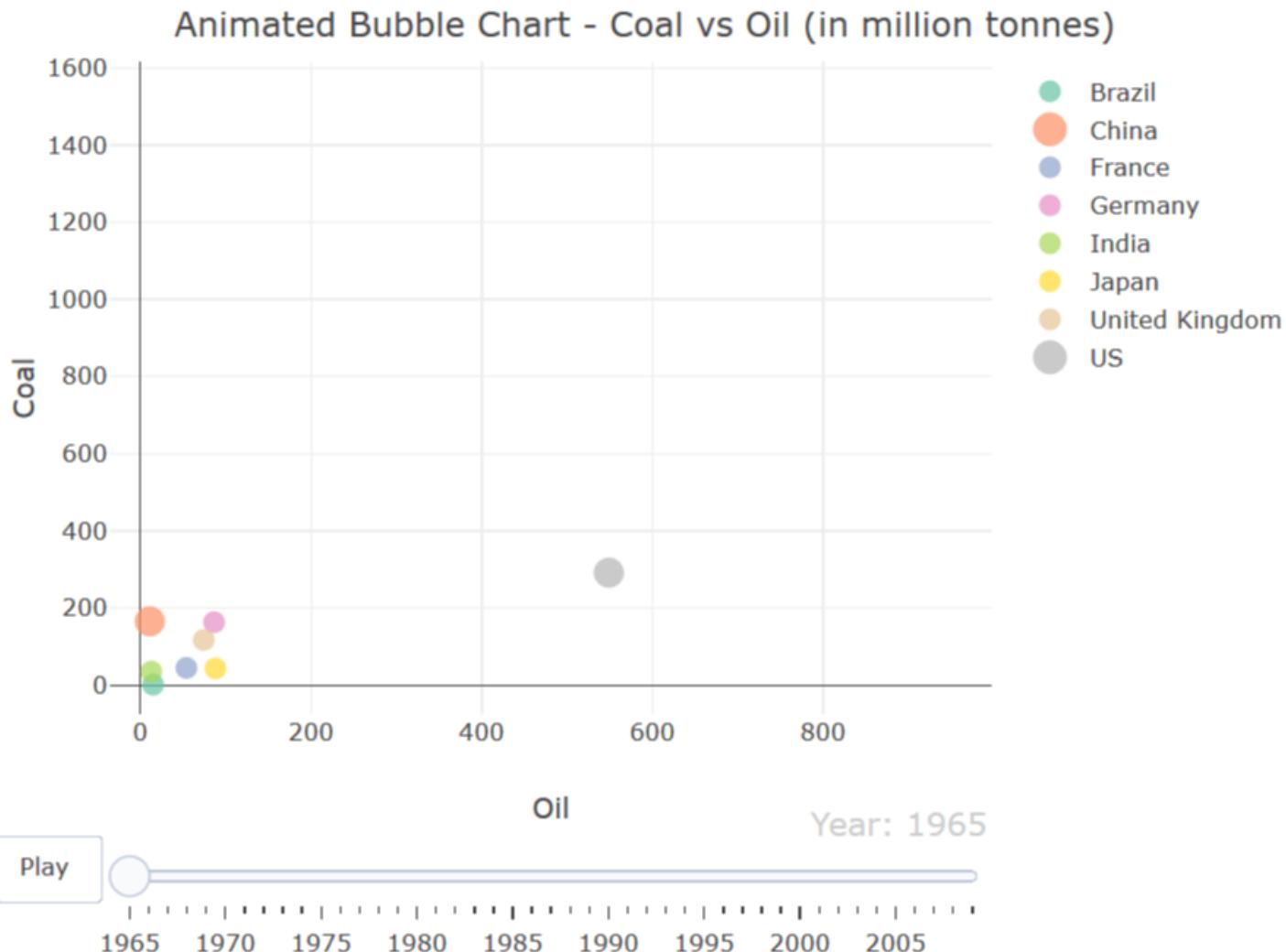
Bubble Chart

Bubble Chart Animated with Cubic Easing

```
oc_data = read.csv2("Oilcoal.csv")

x <- plot_ly(oc_data, x= ~Oil, y= ~Coal, frame = ~Year, color = ~Country) %>%
  add_markers(size = ~Marker.size, marker = list(sizemin = 5)) %>%
  animation_opts(500, easing = "cubic", redraw = F) %>%
  layout(title = "Animated Bubble Chart - Coal vs Oil (in million tonnes)")

knitr:::include_graphics('./6.1.1.png')
```



Bar Chart

Bar Chart Animated with Elastic Easing

```

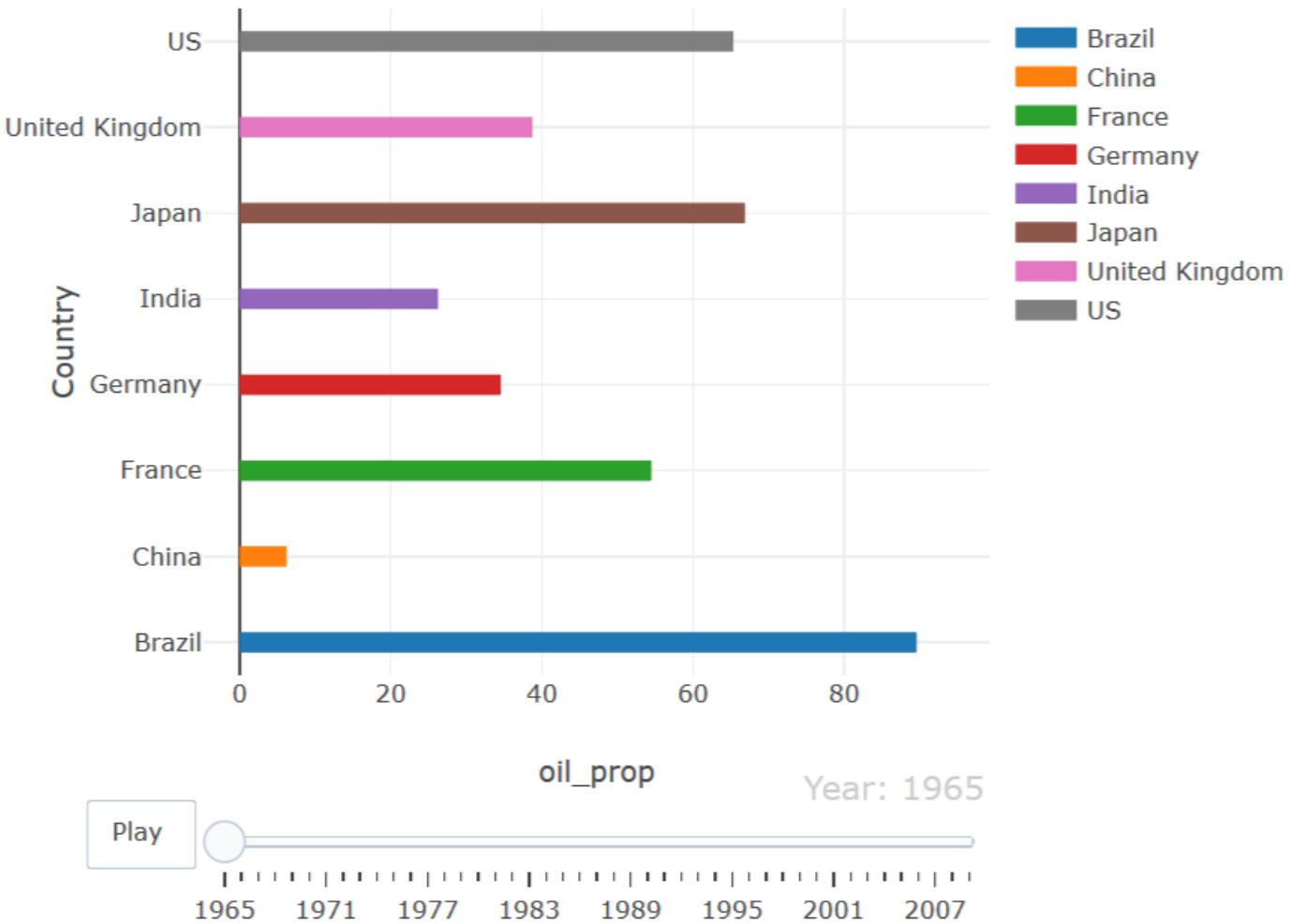
oc_data$oil_prop = 100 * oc_data$Oil / (oc_data$Oil + oc_data$Coal)
oc_data_zero = oc_data
oc_data_zero$oil_prop = 0
oc_line_anim_data = rbind(oc_data, oc_data_zero)

x <- plot_ly(oc_line_anim_data, x = ~oil_prop, y = ~Country, frame = ~Year) %>%
  add_lines(split = ~Country, line = list(width = 10)) %>%
  animation_opts(500, easing = "elastic", redraw = F) %>%
  layout(title = "Animated Bar Chart - Percentage Fuel Consumption of Oil")

knitr:::include_graphics('./6.2.1.png')

```

Animated Bar Chart - Percentage Fuel Consumption of Oil



Tour and Projection

Animated with grand tour

```
oc_tour_data = cast(oc_data[,1:3], Year ~ Country, value = "Coal")

mat <- rescale(oc_tour_data)
set.seed(12345)
#tour <- new_tour(mat, grand_tour(), NULL)
tour<- new_tour(mat, guided_tour(cmass), NULL)

steps <- c(0, rep(1/15, 200))
Projs<-lapply(steps, function(step_size){
  step <- tour(step_size)
  if(is.null(step)) {
    .GlobalEnv$tour<- new_tour(mat, guided_tour(cmass), NULL)
    step <- tour(step_size)
```

```

    }
    step
})

# projection of each observation
tour_dat <- function(i) {
  step <- Projs[[i]]
  proj <- center(mat %*% step$proj)
  data.frame(x = proj[,1], y = proj[,2], state = rownames(mat))
}

# projection of each variable's axis
proj_dat <- function(i) {
  step <- Projs[[i]]
  data.frame(
    x = step$proj[,1], y = step$proj[,2], variable = colnames(mat)
  )
}

stepz <- cumsum(steps)

# tidy version of tour data

tour_dats <- lapply(1:length(steps), tour_dat)
tour_datz <- Map(function(x, y) cbind(x, step = y), tour_dats, stepz)
tour_dat <- dplyr::bind_rows(tour_datz)

# tidy version of tour projection data
proj_dats <- lapply(1:length(steps), proj_dat)
proj_datz <- Map(function(x, y) cbind(x, step = y), proj_dats, stepz)
proj_dat <- dplyr::bind_rows(proj_datz)

ax <- list(
  title = "", showticklabels = FALSE,
  zeroline = FALSE, showgrid = FALSE,
  range = c(-1.1, 1.1)
)

# for nicely formatted slider labels
options(digits = 3)
tour_dat <- highlight_key(tour_dat, ~state, group = "A")

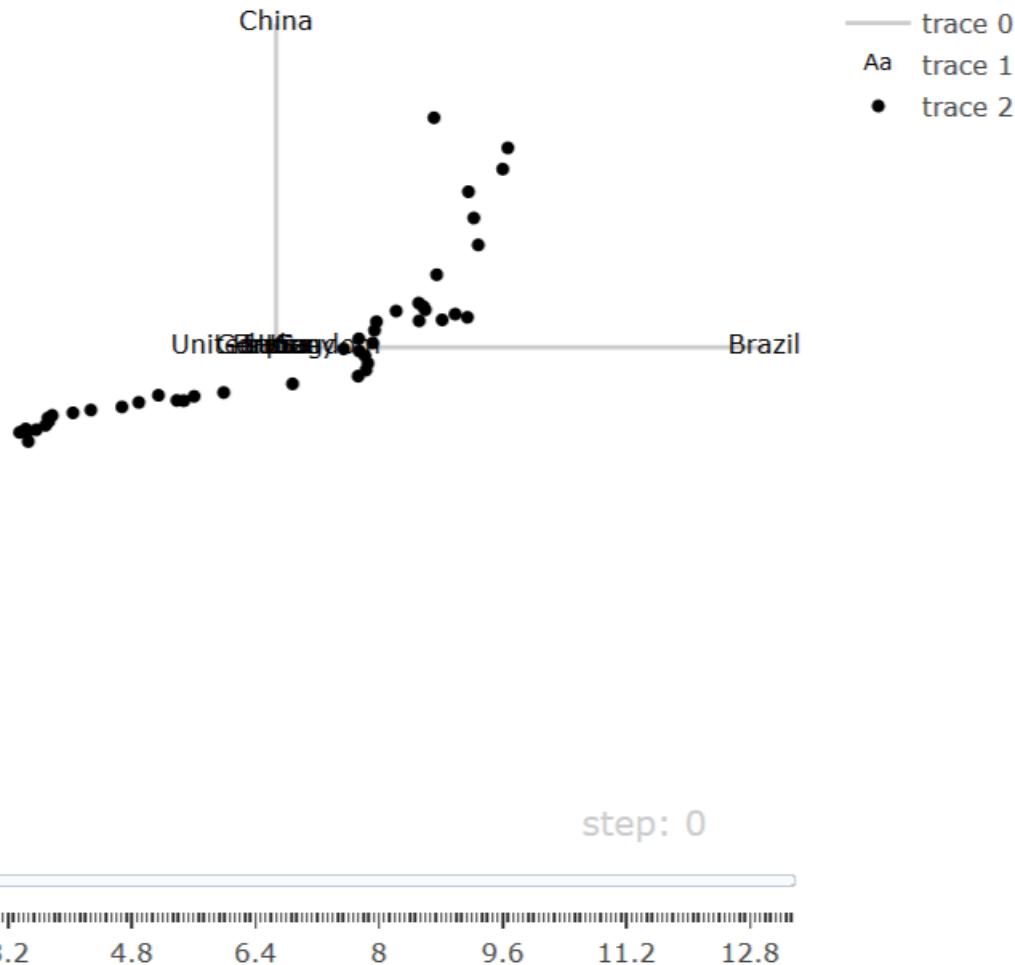
# step = 7.67

x <- proj_dat %>%
  plot_ly(x = ~x, y = ~y, frame = ~step, color = I("black")) %>%
  add_segments(xend = 0, yend = 0, color = I("gray80")) %>%
  add_text(text = ~variable) %>%
  add_markers(data = tour_dat, text = ~state, ids = ~state, hoverinfo = "text") %>%
  layout(xaxis = ax, yaxis = ax, title = "Guided 2D tour visualizing coal consumption of countries")

knitr::include_graphics('./7.1.1.png')

```

Guided 2D tour visualizing coal consumption of countries

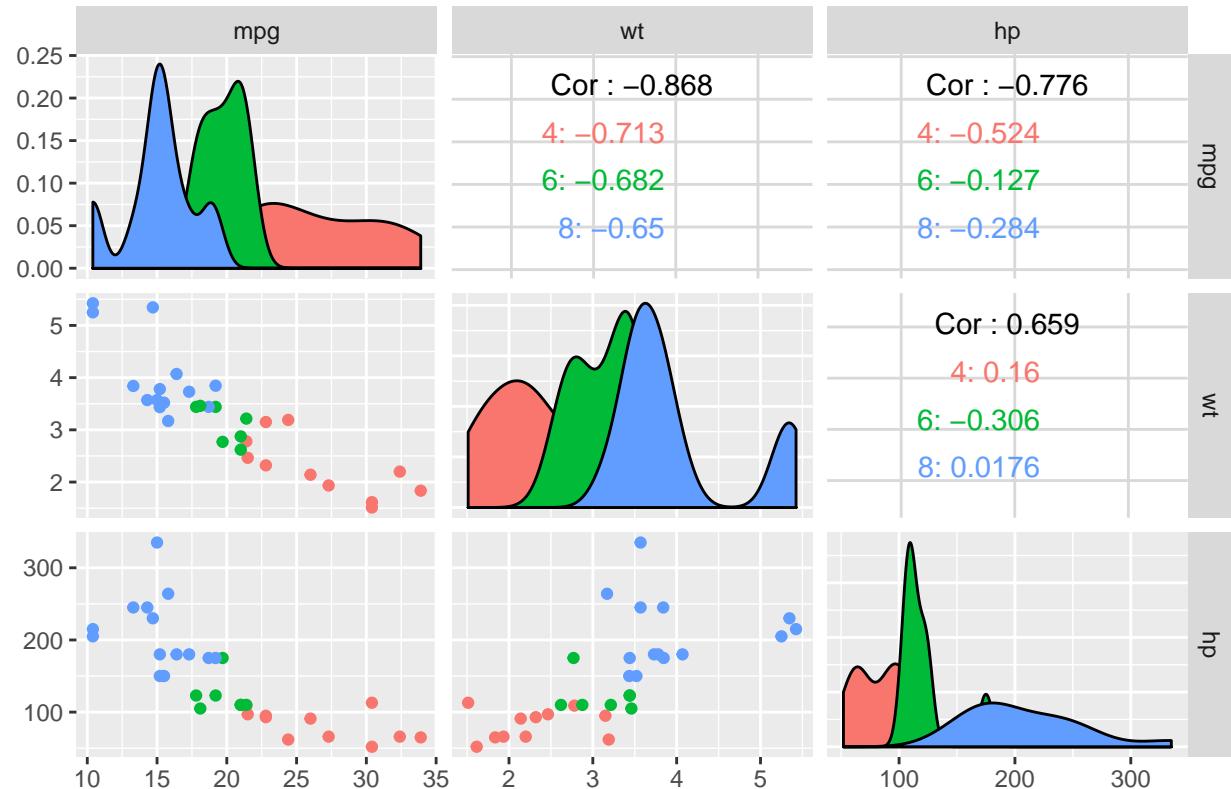


Multiple Plots

QC Scatter Matrix Plots using GGally library

```
ggpairs(mtcars, columns = c("mpg", "wt", "hp"),
        title = "Pair wise cars data vs. cluster(cyl)",
        ggplot2::aes(colour=as.factor(cyl)))
```

Pair wise cars data vs. cluster(cyl)



Density Plots

Density Plot with Outlier Highlight

```
plot_density_with_outliers <- function(var_data, col_name){
  p <- NULL
  df_data = setNames(data.frame(var_data), col_name)
  if(length(get_outliers(df_data[[col_name]])) > 0){
    p <- ggplot(df_data) +
      geom_density(aes_string(x=col_name), fill = "lightblue", color = "darkblue") +
      geom_point(data=df_data[get_outliers(df_data[[col_name]]),,drop=FALSE],
                 aes_string(x=col_name, y=0, shape=23, size=2, colour="black", fill="red"))
  }
  else{
    p <- ggplot(df_data) +
      ggtitle(paste("n"))
      theme(plot.title = element_text(hjust = 0.5)) +
      geom_density(aes_string(x=col_name), fill = "lightblue", color = "darkblue")
  }

  return(p)
}

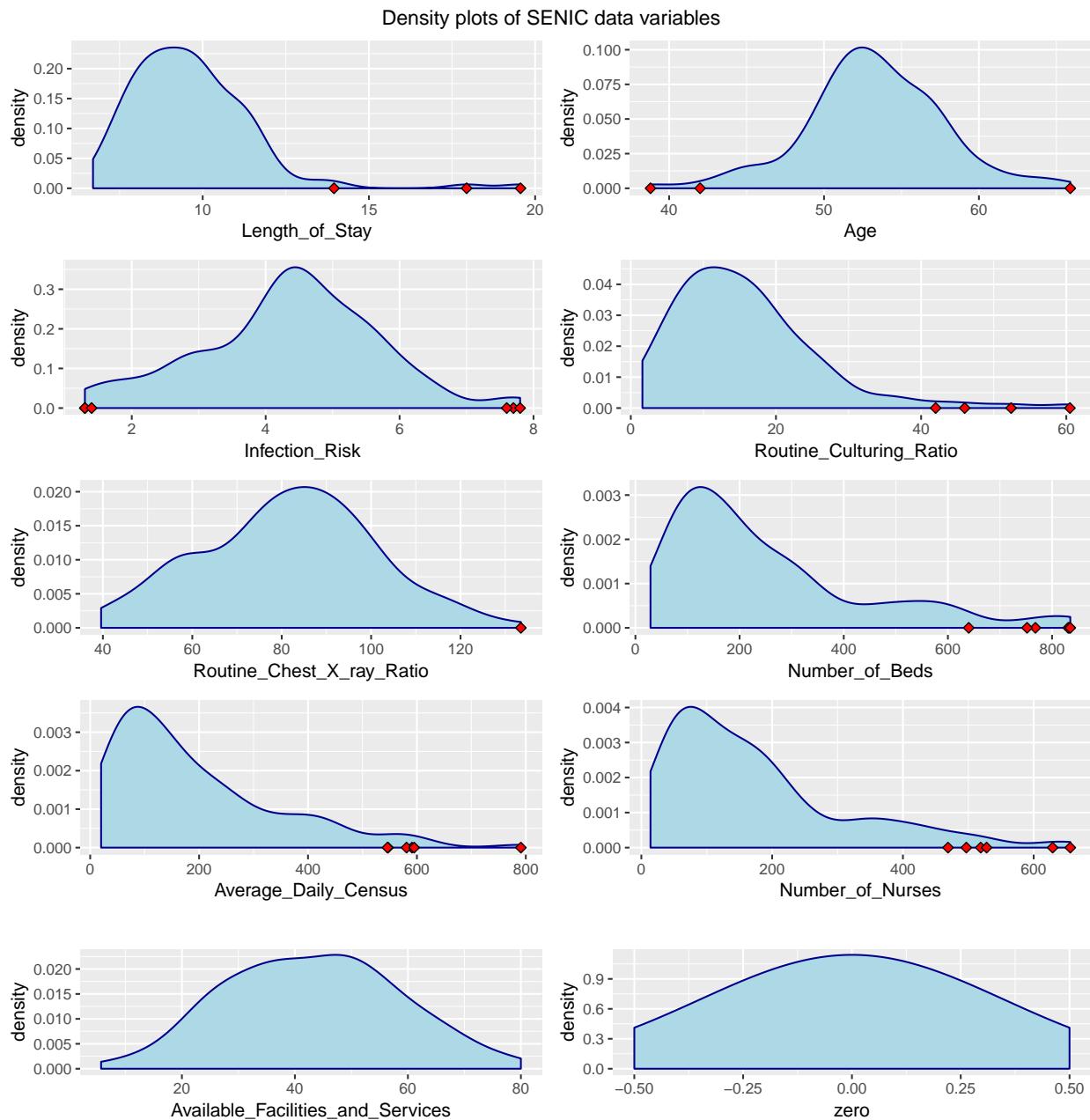
categorical_columns = c("Medical_School_Affiliation", "Region")
```

```

ID_columns = c("Identification_Number")
quantitative_columns = setdiff(colnames(senic_data), c(categorical_columns, ID_columns))

plot_list = mapply(plot_density_with_outliers, senic_data[, quantitative_columns],
                   colnames(senic_data[, quantitative_columns]), SIMPLIFY = FALSE)
plot_matrix <- arrangeGrob(grobs = plot_list, ncol = 2)
grid.arrange(plot_matrix, respect=TRUE, top="Density plots of SENIC data variables")

```



Shiny

```
#UI component
ui <- fluidPage(
  sliderInput(inputId="bw_value", label="Choose bandwidth size", value=4.5, min=0.1, max=80),
  checkboxGroupInput("selected_variables", "Variables to show: ", quantitative_columns, inline=TRUE),
  plotOutput("densPlot", height = "650px")
)

plot_density_with_outliers_shiny <- function(df_data, col_name, bw){
  p <- NULL
  if(length(get_outliers(senic_data[[col_name]])) > 0){
    p <- ggplot(df_data) +
      ggtitle(paste("Density plot of ", col_name)) +
      theme(plot.title = element_text(hjust = 0.5)) +
      geom_density(aes_string(x=col_name), fill = "lightblue", color = "darkblue", bw=bw) +
      geom_point(data=df_data[get_outliers(df_data[[col_name]]),],
                 aes_string(x=col_name, y=0), shape=23, size=2, colour="black", fill="red")
  }
  else{
    p <- ggplot(df_data) +
      ggtitle(paste("Density plot of ", col_name)) +
      theme(plot.title = element_text(hjust = 0.5)) +
      geom_density(aes_string(x=col_name), fill = "lightblue", color = "darkblue", bw=bw)
  }

  return(p)
}

server <- function(input, output) {

  output$densPlot <- renderPlot({

    selected_columns = input$selected_variables
    plot_list = vector("list", length(selected_columns))

    if(length(selected_columns) > 0){
      for(i in 1:length(selected_columns)){
        plot_list[[i]] = plot_density_with_outliers_shiny(senic_data, selected_columns[i],
                                                          bw = input$bw_value)
      }
      plot_matrix <- arrangeGrob(grobs = plot_list, ncol = 2)
      grid.arrange(plot_matrix)
    }

  })
}

shinyApp(ui = ui, server = server, options = list(width="800px", height="900px"))
```

Appendix Code

```
# ````{r, ref.label=knitr::all_labels(), echo=TRUE, eval=FALSE}
# ````
```