Untitled Tata iles 2023-06-12 library(reticulate) use_python("C:/Users/tatai/AppData/Local/Programs/Python/Python311/python.exe") Libraries import pandas as pd import numpy as np import seaborn as sns import matplotlib.pyplot as plt sns.set_theme(style = "darkgrid") Data data = pd.read_csv("D:/Downloads/archive (1)/Mall_Customers.csv") data = py\$datadata %>% head() ## CustomerID Gender Age Annual Income (k\$) Spending Score (1-100) ## 1 1 Male 19 15 39 15 16 16 17 17 ## 2 2 Male 21 81 3 Female 20 6 ## 3 ## 4 4 Female 23 77 40 ## 5 5 Female 31 ## 6 6 Female 22 17 76 **EDA** data.info() ## <class 'pandas.core.frame.DataFrame'> ## RangeIndex: 200 entries, 0 to 199 ## Data columns (total 5 columns): ## # Column Non-Null Count Dtype ## --- ----## 0 CustomerID 200 non-null int64 ## 1 Gender 200 non-null object ## 2 Age 200 non-null int64 ## 3 Annual Income (k\$) 200 non-null int64 ## 4 Spending Score (1-100) 200 non-null ## dtypes: int64(4), object(1) ## memory usage: 7.9+ KB data %>% summary() ## CustomerID Gender Annual Income (k\$) Min. :18.00 Min. : 15.00 ## Min. : 1.00 Length:200 ## 1st Qu.: 50.75 Class :character 1st Qu.:28.75 1st Qu.: 41.50 ## Median :100.50 Mode :character Median :36.00 Median : 61.50 ## Mean :100.50 Mean :38.85 Mean : 60.56 ## 3rd Qu.:150.25 3rd Qu.:49.00 3rd Qu.: 78.00 ## Max. :200.00 Max. :70.00 Max. :137.00 ## Spending Score (1-100) ## Min. : 1.00 ## 1st Qu.:34.75 ## Median :50.00 ## Mean :50.20 ## 3rd Qu.:73.00 ## Max. :99.00 colnames(data) <- c("id", "sex", "age", "annuel_score", "spending_score")</pre> data <- data %>% select(-1) data %>% count(sex) %>% ggplot(aes(n,sex,fill=sex)) + geom_col() + scale_fill_grey() + theme_light() Male sex Male Female 60 90 for i in data.columns[2:] : sns.displot(x=i, data = data) plt.title(i) plt.show(); Age 40 30 10 0 30 20 40 50 60 70 Age Annual Income (k\$) 35 30 25 Count 20 15 10 5 120 20 40 60 80 100 140 Annual Income (k\$) Spending Score (1-100) 40 30 Count 20 10 0 20 60 80 100 Spending Score (1-100) Clustering Recipe t_sne_rec <- recipe(~.,data = data) %>% step_normalize(all_numeric()) %>% step_dummy(all_nominal()) df <- t_sne_rec %>% prep(data) %>% juice() **Cross-Validation** $cv \leftarrow vfold_cv(data, v = 5)$ Tuning grid grid <- tibble(num_clusters = seq(1,15))</pre> K-means workflow k_means_spec <- k_means(num_clusters = tune()) %>% set_engine("stats") kmeans_workflow <- workflow() %>% add_model(k_means_spec) %>% add_recipe(t_sne_rec) Tuning tune_res <- tune_cluster(object = kmeans_workflow,resamples = cv,grid = grid)</pre> tune_res %>% autoplot() sse_total 516.425 516.400 -516.375 516.350 sse_within_total 500 -400 -300 -200 -100 -12 Elbow method (5 clusters) final_kmeans <- kmeans_workflow %>% update_model(k_means_spec %>% set_args(num_clusters = 5)) %>% fit(data) augment(final_kmeans, new_data = data) %>% $plot_ly(x = \sim plot_ly(x = \sim$

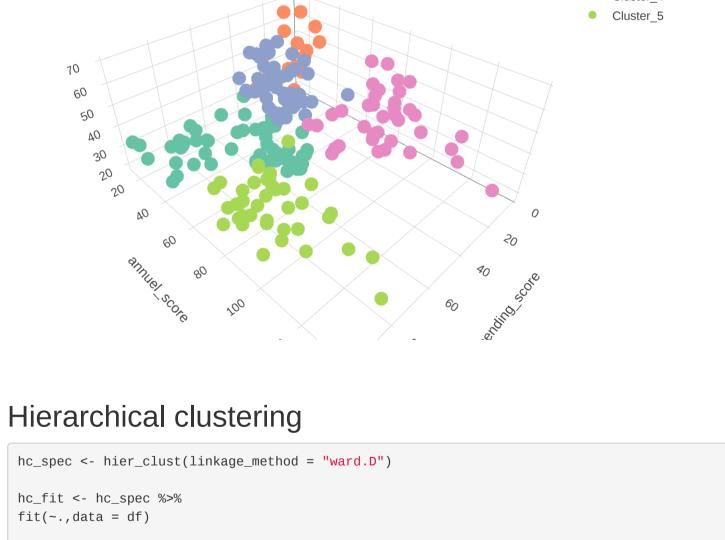
3D representation

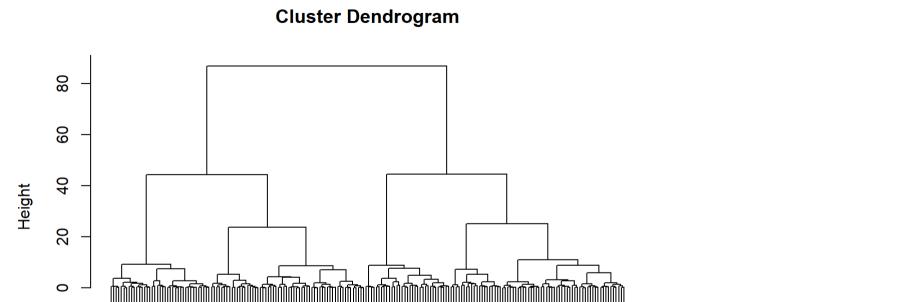
hc_fit\$fit %>% plot()

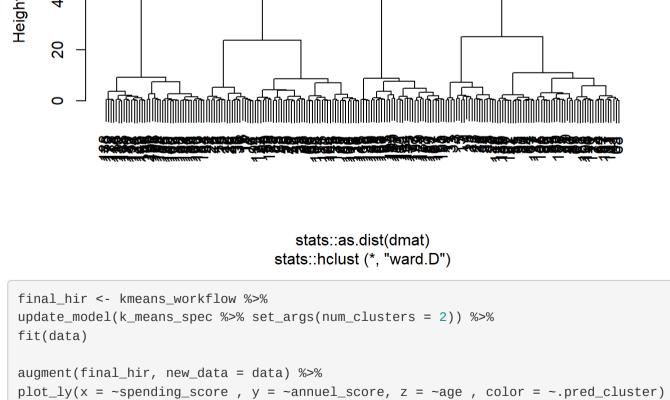
No trace type specified: ## Based on info supplied, a 'scatter3d' trace seems appropriate.

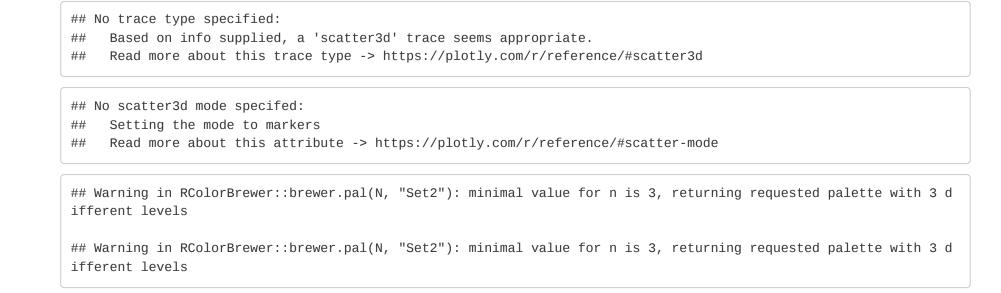
Read more about this trace type -> https://plotly.com/r/reference/#scatter3d

No scatter3d mode specifed: ## Setting the mode to markers ## Read more about this attribute -> https://plotly.com/r/reference/#scatter-mode Cluster_1 Cluster_2 Cluster_3 Cluster_4 Cluster_5









Cluster_1 Cluster_2