



Data Science Project - 17

STUDENT'S NAME

Kareem Mohamed abdelghny elramady
2406247728

Amr Mohamed Shehata Mohamed Saleh
2406246457

Marwan Mohammed Shehata Ahmad
2406249854

Mazen amr Mahmoud fouad
2406248857

The problem

Businesses collect a large amount of data related to customers, including transaction histories, details about the customers. However, without proper analysis, it is challenging to extract meaningful insights from this data.





The role of each member

Mazen:

- Read data
- Clean data
- Visualisation

Marwan:

- K-means

Amr:

- Association

Kareem:

- GUI (Shiny)

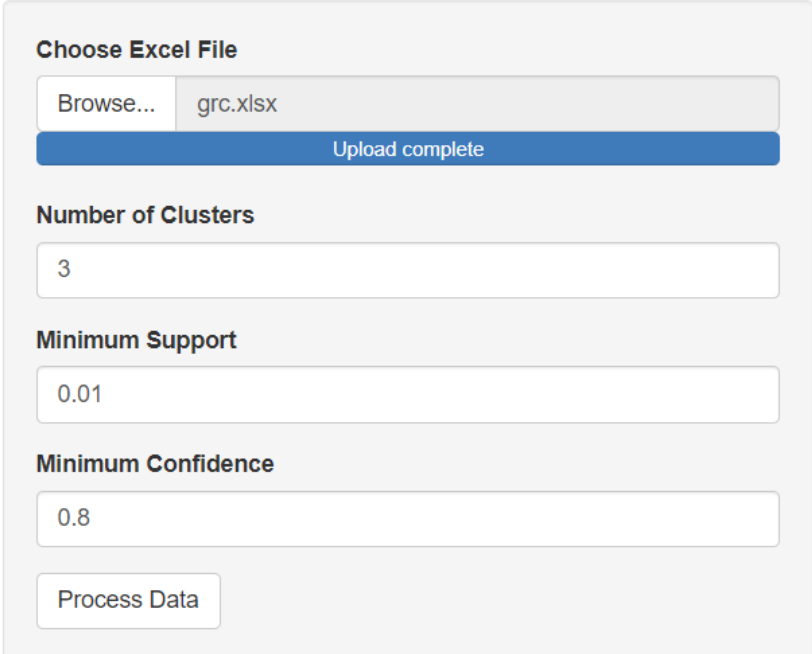
What will the program do ?

The program takes a huge data set and turns it into numerous benefits which are:

1. **Grouping Customers:** Divides customers by spending and age to create better marketing plans.
2. **Finding Patterns:** Identifies which products are often bought together to increase sales.
3. **Making Better Choices:** Helps businesses plan based on customer needs and spending habits.

Input of the program

1. Excel file
2. Number of clusters
3. Minimum support
4. Minimum confidence



The screenshot displays the user interface for the program's input. It features a 'Choose Excel File' section with a 'Browse...' button and a file named 'grc.xlsx'. Below this is a blue bar indicating 'Upload complete'. The 'Number of Clusters' is set to 3. The 'Minimum Support' is set to 0.01. The 'Minimum Confidence' is set to 0.8. A 'Process Data' button is located at the bottom.

Input Parameter	Value
Excel File	grc.xlsx
Number of Clusters	3
Minimum Support	0.01
Minimum Confidence	0.8

Output of the program

1. Graphs



2. Cluster results

Graphs Cluster Results Association Rules

Show 10 entries Search:

	customer	age	total	Cluster
1	Farida	22	794570	1
2	Hanan	22	818543	2
3	Eman	23	772871	1
4	Mohamed	25	930510	3
5	Walaa	29	900797	3
6	Ahmed	30	829587	2
7	Sameh	35	869668	3
8	Magdy	36	901010	3
9	Rania	37	893789	3
10	Sayed	37	820900	2

Showing 1 to 10 of 15 entries

Previous 1 2 Next

3. Association Rules

Graphs

Cluster Results

Association Rules

Show

10

entries

Search:

rules	support	confidence	coverage	lift	count
1 {items=rolls/buns} => {count=[1,2]}	0.01108512152954337	1	0.01108512152954337	4.558646267964765	109
2 {items=bottled beer} => {count=[1,2]}	0.01220380351876335	1	0.01220380351876335	4.558646267964765	120
3 {items=whole milk} => {count=[1,2]}	0.01230550188141971	1	0.01230550188141971	4.558646267964765	121
4 {items=soda} => {count=[1,2]}	0.01576324621173599	1	0.01576324621173599	4.558646267964765	155
5 {items=canned beer} => {count=[1,2]}	0.02633987592799756	1	0.02633987592799756	4.558646267964765	259
6 {customer=Sayed} => {city=Cairo}	0.06284958812163124	1	0.06284958812163124	5.045151359671626	618
7 {customer=Sayed} => {rnd=[10,15]}	0.06284958812163124	1	0.06284958812163124	2.52906378600823	618
8 {customer=Sayed} => {age=[37,60]}	0.06284958812163124	1	0.06284958812163124	2.165382074432944	618
9 {customer=Ahmed} => {city=Giza}	0.06335807993491305	1	0.06335807993491305	15.78330658105939	623
10 {city=Giza} => {customer=Ahmed}	0.06335807993491305	1	0.06335807993491305	15.78330658105939	623

Showing 1 to 10 of 4,975 entries

Previous

1

2

3

4

5

...

498

Next

The full description of the dataset

This dataset is an organized data which illustrates the customer's movements among the grocery.

In which in shows (items, total payment, customer names, age, city, payment type).

Projects steps

- Libraries used

```
1 ##### libraries
2 library(shiny)
3 library(readxl)
4 library(arules)
5 library(ggplot2)
6 library(DT)
7
```

- Build the main structure of shiny code,

And using “ title panel() ” to insert the title “ data science project ”.

```
10 ##### ui
11 ui <- fluidPage (
12   titlePanel("Data Science Project"),
13
```

- And write “ sidebarlayout() ” and use attributes (fileInput, numericInput) to allow users to insert data.

```
14 sidebarLayout(
15   sidebarPanel(
16     fileInput("file", "Choose Excel File", accept = ".xlsx"),
17     numericInput("n_clusters", "Number of Clusters", value = 3, min = 1),
18     numericInput("min_support", "Minimum Support", value = 0.01, min = 0, max = 1, step = 0.01),
19     numericInput("min_confidence", "Minimum Confidence", value = 0.8, min = 0, max = 1, step = 0.01),
20     actionButton("process_btn", "Process Data")
21   ),
22
```

- And use “ mainPanel ” that shows us “The graphs, Cluster results, Association rules”.

```
23 mainPanel(  
24   tabsetPanel(  
25     tabPanel("Graphs",  
26       fluidRow(  
27         column(6, plotOutput("payment_plot")),  
28         column(6, plotOutput("age_plot"))  
29       ),  
30       fluidRow(  
31         column(6, plotOutput("city_plot")),  
32         column(6, plotOutput("hist_plot"))  
33       )  
34     ),  
35     tabPanel("Cluster Results", DTOutput("cluster_table")),  
36     tabPanel("Association Rules", DTOutput("assoc_rules"))  
37   )  
38 )  
39 )  
40 )  
41
```


- Declare variable “data, processed_data, assoc_rules”,
And make sure that data not empty.

```
45 ##### server logic
46 server <- function(input, output, session) {
47
48   # Temporary data recording
49   data <- reactiveVal(NULL)
50   processed_data <- reactiveVal(NULL)
51   assoc_rules <- reactiveVal(NULL)
52
53
54   observeEvent(input$file, {
55     req(input$file)
56     data(read_xlsx(input$file$datapath))
57   })
58
```

- When pressing the button “process_btn”, the app check if data is empty or not, and clean the data.

```
59
60 observeEvent(input$process_btn, {
61   req(data())
62   cleaned_data <- data()
63   cleaned_data <- unique(cleaned_data)
64   cleaned_data <- na.omit(cleaned_data)
65
```

- Implement K-means algorithm to compare between total payment with customer and age.
And use cluster number taken from user.

```

65 ##### K-means clustering
66 aggregated_data <- aggregate(total ~ customer + age, data = cleaned_data, sum)
67 kmeans_result <- kmeans(aggregated_data[, c("age", "total")], centers = input$n_clusters)
68 aggregated_data$cluster <- kmeans_result$cluster
69 processed_data(aggregated_data)
70
71
72 # view K-means clustering
73 output$cluster_table <- renderDT({ req(processed_data()); datatable(processed_data()) })
74
75

```

- Implement association rule and take minimum support and minimum confidence from user
And convert the data into transactions type and apply association algorithm .

```

77 ##### Association Rule Mining
78 transactions <- as(cleaned_data, "transactions")
79 rules <- apriori(transactions, parameter = list(supp = input$min_support, conf = input$min_confidence))
80 assoc_rules(rules)
81
82 # view association rules
83 output$assoc_rules <- renderDT({ req(assoc_rules()); datatable(as(assoc_rules(), "data.frame")) })
84 })
85
86

```

- This graph shows a pie chart for a comparison between total of cash and credit .

```

87 ##### graphs
88 # pie chart ==> total ~ payment type
89 output$payment_plot <- renderPlot({
90   req(data())
91   payment_totals <- aggregate(total ~ paymentType, data = data(), sum)
92   pie(payment_totals$total, labels = paste(payment_totals$paymentType, "\n",
93     round(100 * payment_totals$total / sum(payment_totals$total), 1), "%"),
94     col = rainbow(length(payment_totals$paymentType)), main = "Total Payments by Payment Type")
95 })
96
97
98 # Line Chart ==> total ~ age

```

- This graph shows a line chart for a comparison between total and age .

```

98 # Line Charts ==> total ~ age
99 output$age_plot <- renderPlot({
100   req(data())
101   age_totals <- aggregate(total ~ age, data = data(), sum)
102   plot(age_totals$age, age_totals$total, type = "o", col = "blue", main = "Total Payments by Age", xlab = "Age", ylab = "Total Amount")
103 })
104

```

- This graph shows a bar plot for a comparison between total and city .

```

105 # bar plot ==> total ~ city
106 output$city_plot <- renderPlot({
107   req(data())
108   city_totals <- aggregate(total ~ city, data = data(), sum)
109   city_totals <- city_totals[order(-city_totals$total), ]
110   barplot(city_totals$total, names.arg = city_totals$city, col = rainbow(length(city_totals$city)),
111     main = "Total Payments by City (Descending)", las = 2, xlab = "City", ylab = "Total Amount")
112 })
113

```

- This graph shows a hist plot for distribution of total spending .

```

114 # hist plot ==> Distribution of Total Spending
115 output$hist_plot <- renderPlot({
116   req(data())
117   hist(data()$total, breaks = 20, col = "skyblue", main = "Distribution of Total Spending",
118     xlab = "Total Spending", ylab = "Frequency", border = "white")
119 })
120 }
121

```

- Open app using the function shiny app.

```

122
123 # open the app
124 shinyApp(ui = ui, server = server)

```