R. Notebook

```
# Load required libraries
library(tidyverse)
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr
           1.1.4 v readr
                                    2.1.5
## v forcats 1.0.0
                                    1.5.1
                        v stringr
## v ggplot2 3.5.1
                        v tibble
                                    3.2.1
## v lubridate 1.9.4
                       v tidyr
                                    1.3.1
## v purrr
              1.0.4
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                    masks stats::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
library(openxlsx)
library(lubridate)
# 1. Read the CSV file
customer_data <- read.csv("customer_sales.csv")</pre>
# 2. Clean and analyze the data
# Convert purchase_date to Date type
customer_data$purchase_date <- as.Date(customer_data$purchase_date)</pre>
# Check for missing values
missing_values <- colSums(is.na(customer_data))</pre>
print("Missing values in each column:")
## [1] "Missing values in each column:"
print(missing_values)
##
       customer_id
                      purchase_date
                                              amount product_category
##
                 0
                                                   0
                             gender
##
      customer_age
                                            location
# Basic data summary
summary_stats <- summary(customer_data)</pre>
print("Data summary:")
## [1] "Data summary:"
print(summary_stats)
##
    customer_id
                  purchase_date
                                           amount
                                                        product_category
## Min.
          :1001 Min.
                        :2023-01-05
                                       Min. : 22.50 Length:30
## 1st Qu.:1003 1st Qu.:2023-01-25
                                       1st Qu.: 48.38
                                                        Class : character
## Median :1006 Median :2023-02-18
                                       Median: 79.25 Mode: character
## Mean :1006
                                       Mean : 90.41
                  Mean :2023-02-16
```

```
## 3rd Qu.:1008 3rd Qu.:2023-03-07 3rd Qu.:119.25
## Max. :1010 Max. :2023-03-30 Max. :220.00
                                      location
## customer_age gender
## Min.
         :22.0
                 Length:30
                                     Length:30
## 1st Qu.:31.0
                 Class : character Class : character
## Median :39.0
                 Mode :character Mode :character
## Mean :40.8
## 3rd Qu.:49.0
## Max.
          :65.0
# 3. Segment customers based on spending levels
# Calculate total spending per customer
customer_spending <- customer_data %>%
  group_by(customer_id) %>%
  summarise(
   total_spending = sum(amount),
   average_purchase = mean(amount),
   num_purchases = n(),
   last_purchase = max(purchase_date)
  ) %>%
  mutate(
    spending_category = case_when(
      total_spending >= 300 ~ "High Spender",
     total_spending >= 150 & total_spending < 300 ~ "Medium Spender",
     TRUE ~ "Low Spender"
   )
  )
# Add demographic information to customer spending
customer_demographics <- customer_data %>%
  group_by(customer_id) %>%
  summarise(
   age = first(customer_age),
    gender = first(gender),
   location = first(location)
  )
customer_spending <- customer_spending %>%
 left_join(customer_demographics, by = "customer_id")
# Define age groups
customer_spending <- customer_spending %>%
 mutate(
   age_group = case_when(
     age < 30 \sim "18-29".
      age \geq 30 \& age < 45 ~ "30-44",
     age \geq 45 \& age < 60 ~ "45-59",
      TRUE ~ "60+"
   )
 )
# 4. Create multiple analysis perspectives
# Product category analysis
product_analysis <- customer_data %>%
  group_by(product_category) %>%
```

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summarise(
   total_revenue = sum(amount),
   num_transactions = n(),
   avg_transaction = mean(amount),
   unique_customers = n_distinct(customer_id)
  arrange(desc(total_revenue))
# Age group analysis
age_analysis <- customer_data %>%
  mutate(age_group = case_when(
    customer_age < 30 ~ "18-29",
    customer age \geq 30 \& customer age < 45 \sim "30-44",
    customer_age >= 45 & customer_age < 60 ~ "45-59",</pre>
   TRUE ~ "60+"
  )) %>%
  group_by(age_group, gender) %>%
  summarise(
   total_revenue = sum(amount),
   num_transactions = n(),
   avg_transaction = mean(amount),
   unique_customers = n_distinct(customer_id)
 )
## `summarise()` has grouped output by 'age_group'. You can override using the
## `.groups` argument.
# Location analysis
location_analysis <- customer_data %>%
  group_by(location) %>%
  summarise(
   total_revenue = sum(amount),
   num_transactions = n(),
   avg_transaction = mean(amount),
   unique_customers = n_distinct(customer_id)
 )
# 5. Track monthly sales trend
monthly_sales <- customer_data %>%
  mutate(month = floor_date(purchase_date, "month")) %>%
  group_by(month) %>%
  summarise(
   total_revenue = sum(amount),
   num_transactions = n(),
   avg_transaction = mean(amount),
   unique_customers = n_distinct(customer_id)
 )
# Create marketing impact analysis - simplified approach
marketing_impact <- data.frame(</pre>
  spending_category = c("High Spender", "Medium Spender", "Low Spender")
# Calculate metrics for each spending category
for (category in marketing_impact$spending_category) {
```

```
subset_data <- customer_spending[customer_spending$spending_category == category, ]</pre>
  marketing_impact[marketing_impact$spending_category == category, "num_customers"] <- nrow(subset_data
  marketing_impact[marketing_impact$spending_category == category, "total_revenue"] <- sum(subset_data$
  marketing_impact[marketing_impact$spending_category == category, "avg_spending"] <- mean(subset_data$
}
# Add conversion rates and potential revenue
marketing_impact$conversion_rate <- ifelse(marketing_impact$spending_category == "High Spender", 0.15,
                                   ifelse(marketing_impact$spending_category == "Medium Spender", 0.10,
marketing_impact$avg_additional_purchase <- ifelse(marketing_impact$spending_category == "High Spender"
                                            ifelse(marketing_impact$spending_category == "Medium Spender
marketing_impact$potential_revenue <- marketing_impact$num_customers *
                              marketing_impact$conversion_rate *
                              marketing_impact$avg_additional_purchase
# Create customer recommendations - simplified approach
customer_recommendations <- customer_spending %>%
  select(customer_id, total_spending, spending_category, num_purchases, age_group, gender, location)
# Add recommendation column
customer recommendations $ recommendation <- NA
for (i in 1:nrow(customer_recommendations)) {
  if (customer_recommendations\spending_category[i] == "High Spender") {
    customer recommendations$recommendation[i] <- "Premium loyalty program"</pre>
  } else if (customer_recommendations$spending_category[i] == "Medium Spender" &&
             customer_recommendations$num_purchases[i] > 2) {
    customer_recommendations$recommendation[i] <- "Targeted promotions for most purchased categories"
  } else if (customer_recommendations$spending_category[i] == "Medium Spender") {
    customer_recommendations$recommendation[i] <- "Encourage repeat purchases with discounts"
  } else if (customer_recommendations$spending_category[i] == "Low Spender" &&
             customer_recommendations$num_purchases[i] > 1) {
    customer_recommendations$recommendation[i] <- "Value-based offers"</pre>
  } else {
    customer_recommendations $ recommendation[i] <- "Re-engagement campaign"
}
# Debug print to verify data frames
print("Sample of marketing_impact data:")
## [1] "Sample of marketing_impact data:"
print(head(marketing_impact))
     spending_category num_customers total_revenue avg_spending conversion_rate
##
## 1
         High Spender
                                            1036.23
                                                         345.410
                                                                            0.15
                                   3
## 2
        Medium Spender
                                   6
                                            1557.69
                                                         259.615
                                                                            0.10
## 3
           Low Spender
                                   1
                                             118.24
                                                         118.240
                                                                            0.05
##
   avg_additional_purchase potential_revenue
## 1
                                           90.0
                         200
                         100
                                           60.0
## 2
                          50
## 3
                                           2.5
```

```
print("Sample of customer_recommendations data:")
## [1] "Sample of customer_recommendations data:"
print(head(customer_recommendations))
## # A tibble: 6 x 8
     customer_id total_spending spending_category num_purchases age_group gender
##
##
           <int>
                        <dbl> <chr>
                                                          <int> <chr>
                                                                          <chr>
## 1
           1001
                          361. High Spender
                                                              3 30-44
                                                                          М
## 2
           1002
                          287. Medium Spender
                                                              3 18-29
                                                                          F
## 3
           1003
                          324 High Spender
                                                              3 45-59
                                                                          F
## 4
           1004
                           212. Medium Spender
                                                              3 45-59
                                                                          F
## 5
           1005
                           118. Low Spender
                                                              3 18-29
                                                                          М
## 6
           1006
                           274. Medium Spender
                                                              3 30-44
                                                                          М
## # i 2 more variables: location <chr>, recommendation <chr>
# 6. Export all analyses into a single Excel file with multiple sheets
# Create workbook
wb <- createWorkbook()
# Add worksheets
addWorksheet(wb, "Customer Spending")
addWorksheet(wb, "Product Categories")
addWorksheet(wb, "Age Group Analysis")
addWorksheet(wb, "Location Analysis")
addWorksheet(wb, "Monthly Sales")
addWorksheet(wb, "Marketing Impact")
addWorksheet(wb, "Customer Recommendations")
# Write data to worksheets - with explicit error handling
tryCatch({
 writeData(wb, "Customer Spending", customer_spending)
  print("Wrote Customer Spending worksheet successfully")
}, error = function(e) {
 print(paste("Error writing Customer Spending:", e$message))
## [1] "Wrote Customer Spending worksheet successfully"
tryCatch({
 writeData(wb, "Product Categories", product_analysis)
 print("Wrote Product Categories worksheet successfully")
}, error = function(e) {
  print(paste("Error writing Product Categories:", e$message))
})
## [1] "Wrote Product Categories worksheet successfully"
tryCatch({
 writeData(wb, "Age Group Analysis", age_analysis)
  print("Wrote Age Group Analysis worksheet successfully")
}, error = function(e) {
  print(paste("Error writing Age Group Analysis:", e$message))
```

[1] "Wrote Age Group Analysis worksheet successfully"

```
tryCatch({
  writeData(wb, "Location Analysis", location_analysis)
  print("Wrote Location Analysis worksheet successfully")
}, error = function(e) {
  print(paste("Error writing Location Analysis:", e$message))
## [1] "Wrote Location Analysis worksheet successfully"
tryCatch({
  writeData(wb, "Monthly Sales", monthly_sales)
  print("Wrote Monthly Sales worksheet successfully")
}, error = function(e) {
  print(paste("Error writing Monthly Sales:", e$message))
})
## [1] "Wrote Monthly Sales worksheet successfully"
tryCatch({
 writeData(wb, "Marketing Impact", marketing_impact)
 print("Wrote Marketing Impact worksheet successfully")
}, error = function(e) {
 print(paste("Error writing Marketing Impact:", e$message))
})
## [1] "Wrote Marketing Impact worksheet successfully"
tryCatch({
  writeData(wb, "Customer Recommendations", customer_recommendations)
 print("Wrote Customer Recommendations worksheet successfully")
}, error = function(e) {
 print(paste("Error writing Customer Recommendations:", e$message))
## [1] "Wrote Customer Recommendations worksheet successfully"
# Style the workbook - using a safer approach
for (sheet in names(wb)) {
  # Get number of columns in the sheet's data frame
  cols_count <- switch(sheet,</pre>
                      "Customer Spending" = ncol(customer_spending),
                      "Product Categories" = ncol(product_analysis),
                      "Age Group Analysis" = ncol(age_analysis),
                      "Location Analysis" = ncol(location_analysis),
                      "Monthly Sales" = ncol(monthly_sales),
                      "Marketing Impact" = ncol(marketing_impact),
                      "Customer Recommendations" = ncol(customer_recommendations),
                      1) # Default to 1 if sheet name doesn't match
  # Style headers
  addStyle(wb, sheet, style = createStyle(textDecoration = "bold"), rows = 1, cols = 1:cols_count)
  # Auto-size columns (safely)
  setColWidths(wb, sheet, cols = 1:cols_count, widths = "auto")
# Print worksheet names for verification
```

[1] "Analysis complete! Results exported to: /Users/simransinha/Documents/Semester 3/SDA/Assignments