

Online_Retail_Yash

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```
# loading the package
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.2 --
## v ggplot2 3.3.6      v purrr   0.3.4
## v tibble  3.1.8      v dplyr   1.0.10
## v tidyr   1.2.1      v stringr 1.4.1
## v readr   2.1.2      v forcats 0.5.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
# adding dataset
or <- read.csv("C:/Users/YASH/Downloads/Online_Retail.csv")
```

Q1.

Show the breakdown of the number of transactions by countries i.e., how many transactions are in the dataset for each country (consider all records including cancelled transactions). Show this in total number and also in percentage. Show only countries accounting for more than 1% of the total transactions.

```
#grouped by the countries and showing the countries with more than 1% of total transaction
or_1 <- group_by(or, Country) %>% summarise( Total_Transactions=length(InvoiceNo)) %>% filter(Total_Transactions > 0.01 * length(or_1$InvoiceNo))
or_1
```

```
## # A tibble: 4 x 2
##   Country      Total_Transactions
##   <chr>          <int>
## 1 EIRE              8196
## 2 France            8557
## 3 Germany           9495
## 4 United Kingdom  495478
```

Q2.

Create a new variable 'TransactionValue' that is the product of the existing 'Quantity' and 'UnitPrice' variables. Add this variable to the dataframe.

```
#adding transaction value column
or_2 <- mutate(or,TransactionValue= Quantity * UnitPrice) #>% select(Quantity,UnitPrice,Description) %
head(select(or_2, TransactionValue))
```

```
## TransactionValue
## 1 15.30
## 2 20.34
## 3 22.00
## 4 20.34
## 5 20.34
## 6 15.30
```

Q3.

Using the newly created variable, TransactionValue, show the breakdown of transaction values by countries i.e. how much money in total has been spent each country. Show this in total sum of transaction values. Show only countries with total transaction exceeding 130,000 British Pound.

```
#adding TotalTransactionValue column and filtering it for the transaction greater than 130000
or_3 <- group_by(or_2,Country) %>% summarise(TotalTransactionValue=sum(TransactionValue)) %>% filter(T
or_3
```

```
## # A tibble: 6 x 2
## Country TotalTransactionValue
## <chr> <dbl>
## 1 Australia 137077.
## 2 EIRE 263277.
## 3 France 197404.
## 4 Germany 221698.
## 5 Netherlands 284662.
## 6 United Kingdom 8187806.
```

Q4.

```
Temp=strptime(or$InvoiceDate,format='%m/%d/%Y %H:%M',tz='GMT')
#head(Temp)

or$New_Invoice_Date <- as.Date(Temp)
#or$New_Invoice_Date[20000]- or$New_Invoice_Date[10]
or$Invoice_Day_Week= weekdays(or$New_Invoice_Date)
or$New_Invoice_Hour = as.numeric(format(Temp, "%H"))
or$New_Invoice_Month = as.numeric(format(Temp, "%m"))

# a) Show the percentage of transactions (by numbers) by days of the week

or_4.1 <- group_by(or,Invoice_Day_Week) %>% drop_na() %>% count()
or_4.1$perc <- (or_4.1$n/sum(or_4.1$n)) * 100
or_4.1
```

```
## # A tibble: 6 x 3
## # Groups:   Invoice_Day_Week [6]
##   Invoice_Day_Week      n perc
##   <chr>          <int> <dbl>
## 1 Friday          56127  13.8
## 2 Monday          66382  16.3
## 3 Sunday          63237  15.5
## 4 Thursday        82374  20.2
## 5 Tuesday         68110  16.7
## 6 Wednesday       70599  17.4
```

b) Show the percentage of transactions (by transaction volume) by days of the week

```
or_4.2 <- group_by(or, Invoice_Day_Week) %>% select(Invoice_Day_Week, Quantity) %>% drop_na() %>% count
or_4.2$perc <- (or_4.2$n/sum(or_4.2$n)) * 100
or_4.2
```

```
## # A tibble: 6 x 3
## # Groups:   Invoice_Day_Week [6]
##   Invoice_Day_Week      n perc
##   <chr>          <int> <dbl>
## 1 Friday          82193  15.2
## 2 Monday          95111  17.6
## 3 Sunday          64375  11.9
## 4 Thursday       103857  19.2
## 5 Tuesday        101808  18.8
## 6 Wednesday       94565  17.5
```

c) Show the percentage of transactions (by transaction volume) by month of the year

```
or_4.3 <- group_by(or, New_Invoice_Month) %>% select(New_Invoice_Month, Quantity) %>% drop_na() %>% count
or_4.3$perc <- (or_4.3$n/sum(or_4.3$n)) * 100
or_4.3
```

```
## # A tibble: 12 x 3
## # Groups:   New_Invoice_Month [12]
##   New_Invoice_Month      n perc
##   <dbl> <int> <dbl>
## 1          1 35147  6.49
## 2          2 27707  5.11
## 3          3 36748  6.78
## 4          4 29916  5.52
## 5          5 37030  6.83
## 6          6 36874  6.80
## 7          7 39518  7.29
## 8          8 35284  6.51
## 9          9 50226  9.27
## 10         10 60742 11.2
## 11         11 84711 15.6
## 12         12 68006 12.5
```

```
# d) What was the date with the highest number of transactions from Australia?
or_4.4 <- group_by(or, New_Invoice_Date) %>% drop_na()%>% select(New_Invoice_Date, Country) %>% filter(
max3 <- max(or_4.4$n)
or_4.4.1 <- or_4.4 %>% filter(n == max3)
or_4.4.1
```

```
## # A tibble: 1 x 2
## # Groups:   New_Invoice_Date [1]
##   New_Invoice_Date     n
##   <date>             <int>
## 1 2011-06-15         139
```

Q5.

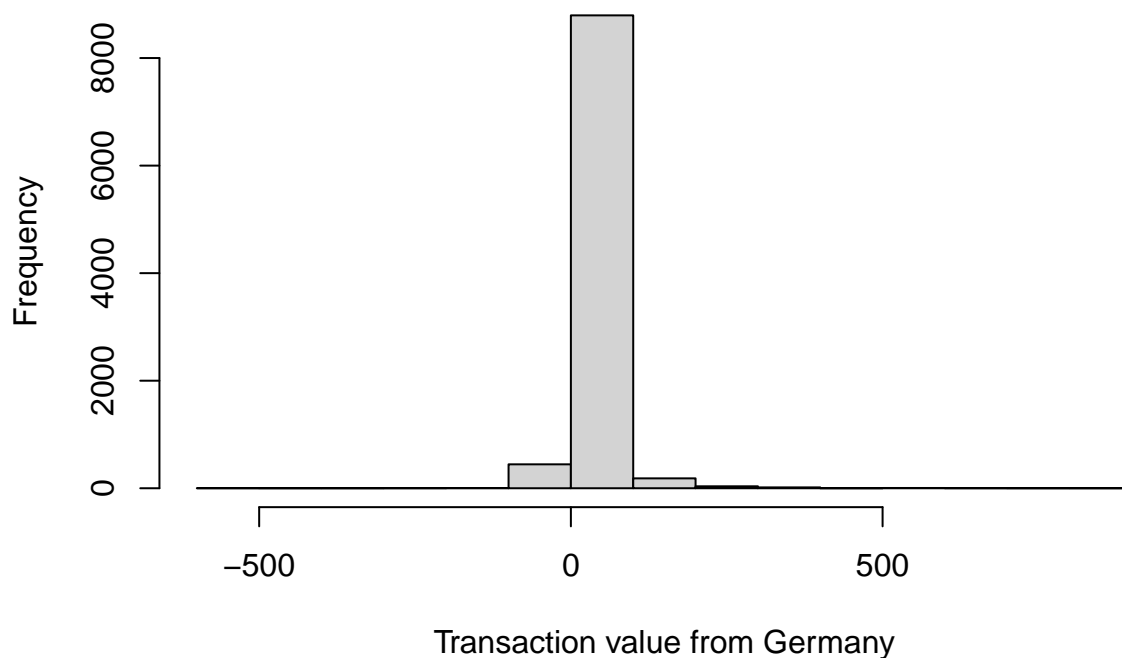
Plot the histogram of transaction values from Germany. Use the hist() function to plot.

```
#filtering or_2 dataset to select Germany as Country and then selecting Country TransactionValue column
or_5 <- filter(or_2, Country == "Germany" ) %>% select(Country, TransactionValue)

#assigning variable for histogram
"Transaction value from Germany" <- or_5$TransactionValue

#creating histogram using hist function
hist(`Transaction value from Germany`)
```

Histogram of Transaction value from Germany



Q6.

1.Which customer had the highest number of transactions? 2.Which customer is most valuable (i.e. highest total sum of transactions)?

```
#1
#counting the total transaction by CustomerID and dropping the observation without ColumnID
or_6.0 <- group_by(or_2, CustomerID) %>% summarise( Total_Transactions=length(TransactionValue)) %>% drop_na()

#finding the largest number
max0 <- max(or_6.0$Total_Transactions)

#filtering the data to get the CustomerID with Maximum transaction
or_6.0_max <- filter(or_6.0, Total_Transactions== max0)

or_6.0_max
```

```
## # A tibble: 1 x 2
##   CustomerID Total_Transactions
##       <int>           <int>
## 1      17841             7983
```

```
#2
#totaling the TotalTransactionValue by CustomerID and dropping the observation without ColumnID
or_6.1 <- group_by(or_2, CustomerID) %>% summarise(TotalTransactionValue=sum(TransactionValue)) %>% drop_na()

#finding the largest number
max1 <- max(or_6.1$TotalTransactionValue)

#filtering the data to get the CustomerID with Maximum TotalTransactionValue
or_6.1_max <- filter(or_6.1, TotalTransactionValue== max1)

or_6.1_max
```

```
## # A tibble: 1 x 2
##   CustomerID TotalTransactionValue
##       <int>           <dbl>
## 1      14646          279489.
```

Q7.

Calculate the percentage of missing values for each variable in the dataset

```
colMeans(is.na(or_2)) #25%(24.92669)
```

```
##      InvoiceNo      StockCode      Description      Quantity
##      0.0000000      0.0000000      0.0000000      0.0000000
##      InvoiceDate      UnitPrice      CustomerID      Country
##      0.0000000      0.0000000      0.2492669      0.0000000
## TransactionValue
##      0.0000000
```

```
#or
```

```
summary(or)
```

```
## InvoiceNo      StockCode      Description      Quantity
## Length:541909 Length:541909 Length:541909 Min. :-80995.00
## Class :character Class :character Class :character 1st Qu.: 1.00
## Mode :character Mode :character Mode :character Median : 3.00
## Mean : 9.55
## 3rd Qu.: 10.00
## Max. : 80995.00
##
## InvoiceDate      UnitPrice      CustomerID      Country
## Length:541909 Min. :-11062.06 Min. :12346 Length:541909
## Class :character 1st Qu.: 1.25 1st Qu.:13953 Class :character
## Mode :character Median : 2.08 Median :15152 Mode :character
## Mean : 4.61 Mean :15288
## 3rd Qu.: 4.13 3rd Qu.:16791
## Max. : 38970.00 Max. :18287
## NA's :135080
## New_Invoice_Date Invoice_Day_Week New_Invoice_Hour New_Invoice_Month
## Min. :2010-12-01 Length:541909 Min. : 6.00 Min. : 1.000
## 1st Qu.:2011-03-28 Class :character 1st Qu.:11.00 1st Qu.: 5.000
## Median :2011-07-19 Mode :character Median :13.00 Median : 8.000
## Mean :2011-07-04 Mean :13.08 Mean : 7.553
## 3rd Qu.:2011-10-19 3rd Qu.:15.00 3rd Qu.:11.000
## Max. :2011-12-09 Max. :20.00 Max. :12.000
##
```

```
#only ColumnID variable has missing value
```

```
or_7 <- (135080*100)/541909
```

```
or_7 # ~25%(24.92669)
```

```
## [1] 24.92669
```

Q8.

What are the number of transactions with missing CustomerID records by countries?

```
#Filtering the Missing CustomeID records and Counting it by Countries under new Column "Total_Transactions"
or_8 <- group_by(or_2,Country ) %>% filter(is.na(CustomerID),Country != "Unspecified") %>% summarise( Total_Transactions = sum(is.na(CustomerID)))
```

```
#First Arranging the Total_Transaction in descending order and adding Percentage column showing Total_Transactions
or_8.1 <- or_8 %>% arrange(desc(Total_Transactions)) %>% mutate(percentage = Total_Transactions / sum(Total_Transactions))
or_8.1
```

```
## # A tibble: 8 x 3
## Country      Total_Transactions percentage
```

##	<chr>	<int>	<dbl>
## 1	United Kingdom	133600	99.1
## 2	EIRE	711	0.527
## 3	Hong Kong	288	0.214
## 4	Switzerland	125	0.0927
## 5	France	66	0.0489
## 6	Israel	47	0.0348
## 7	Portugal	39	0.0289
## 8	Bahrain	2	0.00148

Q9.

On average, how often the costumers comeback to the website for their next shopping? (i.e. what is the average number of days between consecutive shopping)

```
or_9<- or %>% select(CustomerID,New_Invoice_Date) %>% group_by(CustomerID) %>% distinct(New_Invoice_Date)
mean(or_9$Days_Between)
```

```
## Time difference of 38.4875 days
```

Q10.

In the retail sector, it is very important to understand the return rate of the goods purchased by customers. In this example, we can define this quantity, simply, as the ratio of the number of transactions cancelled (regardless of the transaction value) over the total number of transactions. With this definition, what is the return rate for the French customers? Consider the cancelled transactions as those where the ‘Quantity’ variable has a negative value

```
#Return rate for the French customers

#counting the total number of transaction
x.0 <-filter(or_2,Country == "France") %>% count()
#counting the transaction with negative quantity
y.0 <-filter(or_2,Country == "France",Quantity < 0) %>% count()
#finally finding the return rate for the french customers dividing the number of transactions cancelled
or_10 <-summarise(or_2, Return_rate_for_the_French_customers= y.0/ x.0 * 100)
or_10  #(1.741264)
```

```
##           n
## 1 1.741264
```

Q11.

What is the product that has generated the highest revenue for the retailer? (i.e. item with the highest total sum of ‘TransactionValue’).

```

#grouping the dataset by StockCode and using summarise function to find the Total Transaction Value
or_11 <- group_by(or_2, StockCode) %>% summarise( Total_Transac_val = sum(TransactionValue))
#finding the largest Value
max2 <- max(or_11$Total_Transac_val)
#filtering the dataset to get the Stockcode which has the largest Value
or_11.1 <- filter(or_11,Total_Transac_val==max2)
or_11.1

```

```

## # A tibble: 1 x 2
##   StockCode Total_Transac_val
##   <chr>          <dbl>
## 1 DOT            206245.

```

Q12.

How many unique customers are represented in the dataset?

```

#finding the unique customer using the unique() function and dropping the empty observations and then count
or_12 <- data.frame(unique(or_2$CustomerID)) %>% drop_na() %>% count("Total_Unique_customers")
or_12

```

```

##   "Total_Unique_customers"      n
## 1   Total_Unique_customers 4372

```