

Assignment - 2

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```
OnlineRetail <- read.csv("c:/Users/Harika/Downloads/Online_Retail.csv")
```

###The file is loaded into an R DataFrame by the above command.

```
summary(OnlineRetail)
```

```
## 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
```

###The above data represents the summary for the given dataset.

#1 > 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.

```
Country_total_number <- table(OnlineRetail$Country)
Country_total_number
```

```
##
##      Australia      Austria      Bahrain
##      1259          401          19
##      Belgium      Brazil      Canada
##      2069          32          151
##      Channel Islands Cyprus      Czech Republic
##      758          622          30
```

##	Denmark	EIRE	European Community
##	389	8196	61
##	Finland	France	Germany
##	695	8557	9495
##	Greece	Hong Kong	Iceland
##	146	288	182
##	Israel	Italy	Japan
##	297	803	358
##	Lebanon	Lithuania	Malta
##	45	35	127
##	Netherlands	Norway	Poland
##	2371	1086	341
##	Portugal	RSA	Saudi Arabia
##	1519	58	10
##	Singapore	Spain	Sweden
##	229	2533	462
##	Switzerland	United Arab Emirates	United Kingdom
##	2002	68	495478
##	Unspecified	USA	
##	446	291	

###The data above represents the breakdown of the number of transactions by country from the given data.

```
transaction_percent <- round(100*prop.table(Country_total_number),digits = 2)
transaction_percent
```

##			
##	Australia	Austria	Bahrain
##	0.23	0.07	0.00
##	Belgium	Brazil	Canada
##	0.38	0.01	0.03
##	Channel Islands	Cyprus	Czech Republic
##	0.14	0.11	0.01
##	Denmark	EIRE	European Community
##	0.07	1.51	0.01
##	Finland	France	Germany
##	0.13	1.58	1.75
##	Greece	Hong Kong	Iceland
##	0.03	0.05	0.03
##	Israel	Italy	Japan
##	0.05	0.15	0.07
##	Lebanon	Lithuania	Malta
##	0.01	0.01	0.02
##	Netherlands	Norway	Poland
##	0.44	0.20	0.06
##	Portugal	RSA	Saudi Arabia
##	0.28	0.01	0.00
##	Singapore	Spain	Sweden
##	0.04	0.47	0.09
##	Switzerland	United Arab Emirates	United Kingdom
##	0.37	0.01	91.43
##	Unspecified	USA	
##	0.08	0.05	

###The above data represents the percentage of transactions for each country.

```
total <- data.frame(Country=names(Country_total_number),
                    TotalNumber=Country_total_number,
                    percentage=transaction_percent)
total
```

##	Country	TotalNumber.Var1	TotalNumber.Freq
## 1	Australia	Australia	1259
## 2	Austria	Austria	401
## 3	Bahrain	Bahrain	19
## 4	Belgium	Belgium	2069
## 5	Brazil	Brazil	32
## 6	Canada	Canada	151
## 7	Channel Islands	Channel Islands	758
## 8	Cyprus	Cyprus	622
## 9	Czech Republic	Czech Republic	30
## 10	Denmark	Denmark	389
## 11	EIRE	EIRE	8196
## 12	European Community	European Community	61
## 13	Finland	Finland	695
## 14	France	France	8557
## 15	Germany	Germany	9495
## 16	Greece	Greece	146
## 17	Hong Kong	Hong Kong	288
## 18	Iceland	Iceland	182
## 19	Israel	Israel	297
## 20	Italy	Italy	803
## 21	Japan	Japan	358
## 22	Lebanon	Lebanon	45
## 23	Lithuania	Lithuania	35
## 24	Malta	Malta	127
## 25	Netherlands	Netherlands	2371
## 26	Norway	Norway	1086
## 27	Poland	Poland	341
## 28	Portugal	Portugal	1519
## 29	RSA	RSA	58
## 30	Saudi Arabia	Saudi Arabia	10
## 31	Singapore	Singapore	229
## 32	Spain	Spain	2533
## 33	Sweden	Sweden	462
## 34	Switzerland	Switzerland	2002
## 35	United Arab Emirates	United Arab Emirates	68
## 36	United Kingdom	United Kingdom	495478
## 37	Unspecified	Unspecified	446
## 38	USA	USA	291
##	percentage.Var1	percentage.Freq	
## 1	Australia	0.23	
## 2	Austria	0.07	
## 3	Bahrain	0.00	
## 4	Belgium	0.38	
## 5	Brazil	0.01	
## 6	Canada	0.03	

## 7	Channel Islands	0.14
## 8	Cyprus	0.11
## 9	Czech Republic	0.01
## 10	Denmark	0.07
## 11	EIRE	1.51
## 12	European Community	0.01
## 13	Finland	0.13
## 14	France	1.58
## 15	Germany	1.75
## 16	Greece	0.03
## 17	Hong Kong	0.05
## 18	Iceland	0.03
## 19	Israel	0.05
## 20	Italy	0.15
## 21	Japan	0.07
## 22	Lebanon	0.01
## 23	Lithuania	0.01
## 24	Malta	0.02
## 25	Netherlands	0.44
## 26	Norway	0.20
## 27	Poland	0.06
## 28	Portugal	0.28
## 29	RSA	0.01
## 30	Saudi Arabia	0.00
## 31	Singapore	0.04
## 32	Spain	0.47
## 33	Sweden	0.09
## 34	Switzerland	0.37
## 35	United Arab Emirates	0.01
## 36	United Kingdom	91.43
## 37	Unspecified	0.08
## 38	USA	0.05

###The data above combines the total number and percentage of transactions into a table.

```
total <- subset(total,transaction_percent>1)
total
```

##	Country	TotalNumber.Var1	TotalNumber.Freq	percentage.Var1
## 11	EIRE	EIRE	8196	EIRE
## 14	France	France	8557	France
## 15	Germany	Germany	9495	Germany
## 36	United Kingdom	United Kingdom	495478	United Kingdom
##	percentage.Freq			
## 11	1.51			
## 14	1.58			
## 15	1.75			
## 36	91.43			

###The data above represents a subset of the table, showing only countries that account for more than 1.

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

```
library(dplyr)
```

```
##  
## Attaching package: 'dplyr'  
  
## The following objects are masked from 'package:stats':  
##  
## filter, lag  
  
## The following objects are masked from 'package:base':  
##  
## intersect, setdiff, setequal, union
```

```
###This command calls 'dplyr' library.
```

```
OnlineRetail <- OnlineRetail %>% mutate(TransactionValue= Quantity*UnitPrice)  
summary(OnlineRetail$TransactionValue)
```

```
##      Min.      1st Qu.      Median      Mean      3rd Qu.      Max.  
## -168469.60      3.40      9.75      17.99      17.40 168469.60
```

```
###The above data represents the product of the 'Quantity' and 'UnitPrice' variables and assigned the r
```

#3 > 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.

```
data <- summarise(group_by(OnlineRetail,Country),sum_1= sum(TransactionValue))  
Transaction <- filter(data,sum_1 > 130000)  
Transaction
```

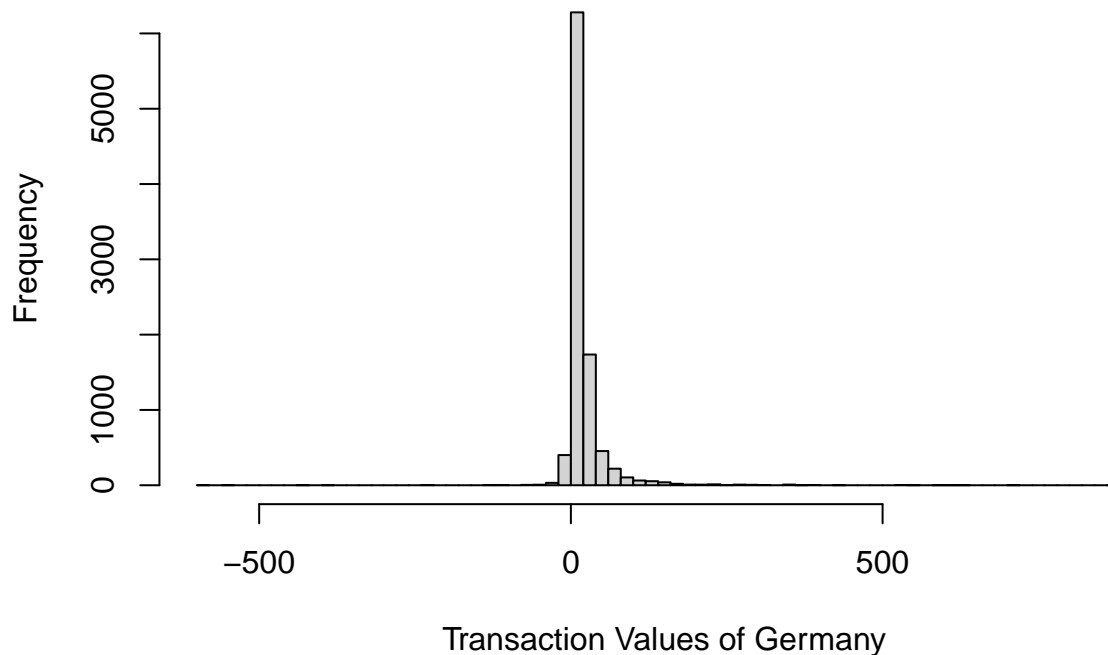
```
## # A tibble: 6 x 2  
##   Country      sum_1  
##   <chr>      <dbl>  
## 1 Australia 137077.  
## 2 EIRE      263277.  
## 3 France    197404.  
## 4 Germany   221698.  
## 5 Netherlands 284662.  
## 6 United Kingdom 8187806.
```

```
###The data above shows the total transaction values for each country. It includes only those countries
```

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

```
Germany_data <- subset(OnlineRetail,Country == "Germany")  
hist(Germany_data$TransactionValue, xlim = c(-600,900),breaks=100, xlab = "Transaction Values of German
```

Histogram of Germany Transaction Values



###This is a representation of the histogram for transaction values in Germany.

#6 > Which customer had the highest number of transactions? Which customer is most valuable (i.e.highest total sum of transactions)?

```
OnlineRetail1 <- na.omit(OnlineRetail)
result1 <- summarise(group_by(OnlineRetail1, CustomerID), sum2 = sum(TransactionValue))
result1[which.max(result1$sum2),]
```

```
## # A tibble: 1 x 2
##   CustomerID    sum2
##       <int>    <dbl>
## 1      14646 279489.
```

```
data2 <- table(OnlineRetail$CustomerID)
data2 <- as.data.frame(data2)
result2 <- data2[which.max(data2$Freq),]
result2
```

```
##           Var1 Freq
## 4043 17841 7983
```

This data represents the customer with the highest number of transactions and the most valuable cus

#7 > Calculate the percentage of missing values for each variable in the dataset.

```
missing_values <- colMeans(is.na(OnlineRetail)*100)
missing_values
```

```
##      InvoiceNo      StockCode      Description      Quantity
##      0.00000      0.00000      0.00000      0.00000
##      InvoiceDate      UnitPrice      CustomerID      Country
##      0.00000      0.00000      24.92669      0.00000
## TransactionValue
##      0.00000
```

###The data above represents Missing Values in the given dataset.

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

```
OnlineRetail2 <- OnlineRetail %>% filter(is.na(CustomerID)) %>% group_by(Country)
summary(OnlineRetail2$Country)
```

```
##      Length      Class      Mode
##      135080 character character
```

###The data above shows the number of transactions with missing CustomerID records, records by countries

#10 > 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.

```
OnlineRetail_Table <- filter(OnlineRetail, Country == "France")
Total_Row <- nrow(OnlineRetail_Table)
Cancel <- nrow(subset(OnlineRetail_Table, TransactionValue < 0))
Cancel
```

```
## [1] 149
```

```
NotCancel <- Total_Row - Cancel
NotCancel
```

```
## [1] 8408
```

```
Return_Rate <- Cancel / Total_Row
Return_Rate
```

```
## [1] 0.01741264
```

###The data above represents values for both cancelled and not cancelled transactions, as well as the r

#11 > What is the product that has generated the highest revenue for the retailer? (i.e. item with the highest total sum of 'TransactionValue')

```
Transaction_Value <- tapply(OnlineRetail$TransactionValue, OnlineRetail$StockCode, sum)
Transaction_Value[which.max(Transaction_Value)]
```

```
##      DOT
## 206245.5
```

###The data above represents the item with the highest total sum of 'Transaction Value'.

#12 > How many unique customers are represented in the dataset? You can use unique() and length() functions.

```
unique_customers <- unique(OnlineRetail$CustomerID)
length(unique_customers)
```

```
## [1] 4373
```

###The above value represents number of unique customers that are present in the given dataset.