Data Science Task

Objectives:

- 1. Performing Exploratory data Analysis.
- 2. Predicting Unit Price Using Prediction Modelling.

Exploratory Data Analysis

• Comparing Quantity of Stock purchased by Country. There are no missing values in the dataset.

Country	Count of	Max of	Min of
Country	Quantity	Quantity	Quantity
0	619	960	-120
1	217	240	-48
2	7	96	2
3	1007	272	-12
4	13	24	2
5	73	288	2
6	351	200	-2
7	296	288	-33
8	21	72	-24
9	211	256	-25
10	3631	1440	-288
11	28	24	1
12	332	144	-2
13	4198	912	-120
14	4731	600	-288
15	71	48	-1
16	84	240	2
17	129	100	-32
18	394	100	-12
19	160	1488	-624
20	25	15	2
21	15	36	6
22	60	45	-2
23	1165	2400	-144
24	537	240	-12
25	157	48	-6
26	756	120	-12
27	32	12	1
28	4	12	-5
29	104	100	-1
30	1256	360	-288
31	213	576	-240
32	887	144	-120
33	135	72	-36
34	33	24	2
35	177917	74215	-80995
36	131	25	1
Grand Total	200000	74215	-80995

The total count of the **quantity of stock sell** is **200000**.

The **top 3** countries with maximum quantity purchased is **74215** with country code **35** followed by country code **23** with purchased quantity is **2400** and country code **19** with purchased quantity **1488**. The top 3 countries with **maximum quantity returned** is country with country code **35** with quantity **-80995** followed by country code **19** with quantity **-624** and country code **30,14** and **10** with quantity **-288** each. Country with code **35** has the highest quantity **sold and return**.

Comparing Unit Price of Stocks by Country.

Country Code	Sum of Unit Price	Average of Unit Price	Max of Unit Price	Min of Unit Price	
0	1681.19	2.715977383	14.95	0	
1	961.8	4.432258065	40	0.36	
2	31.45	4.492857143	9.95	1.25	
3	3562.52	3.53775571	29.95	0.12	
4	55.01	4.231538462	8.25	0.85	
5	171.65	2.351369863	12.75	0.1	
6	1760.73	5.016324786	293	0.19	
7	1991.29	6.727331081	320.69	0.21	
8	70.38	3.351428571	40	0.29	
9	707.96	3.355260664	18	0.29	
10	18703.86	5.15115946	1687.17	0	
11	144.55	5.1625	15	0.55	
12	1574.38	4.742108434	40	0.12	
13	22293.76	5.310566937	4161.06	0	
14	17593.57	3.718784612	599.5	0	
15	459.13	6.466619718	50	0.19	
16	236.01	2.809642857	12.75	0.25	
17	526.75	4.083333333	125	0.06	
18	1690.41	4.290380711	40	0.12	
19	355.92	2.2245	17.55	0.21	
20	151.65	6.066	14.95	0.85	
21	43.45	2.896666667	5.95	1.25	
22	402.23	6.703833333	65	0.19	
23	3113.04	2.672137339	110	0.19	
24	3591.92	6.68886406	700	0	
25	622.14	3.962675159	40	0.19	
26	4582.29	6.061230159	557.72	0.12	
27	150.54	4.704375	14.95	0	
28	9.2	2.3	2.95	1.65	
29	12438.86	119.6044231	2382.92	0.19	
30	6464.97	5.147269108	1715.85	0	
31	822.94	3.863568075	40	0.19	
32	2923.31	3.29572717	40	0.12	
33	311.27	2.305703704	16.95	0.42	
34	115.34	3.495151515	14.95	0.29	
35	580127.274	3.26066241	38970	0	
36	427.06	3.26	16.95	0.29	
Grand Total	690869.804	3.45434902	38970	0	

As, we can observe from the above table the maximum unit price of the stock is 38970 in country with country code 35 and Maximum total unit price is also in the country code 35. Also, the maximum average unit price of the stock is 119.60 that is with the country code 29.

Comparing Quantity of stocks sold and purchased by Year and Quarter.

J &	and Quarter.							
Row Labels	Sum of Quantity Purchased	Count of Quantity Purchased	Max of Quantity Purchased	Min of Quantity Purchased				
<12/1/2010								
<12/1/2010								
Qtr1								
2011	501360	34395	74215	-600				
Qtr2								
2011	495499	39362	4300	-1930				
Qtr3								
2011	627609	47239	3186	-756				
Qtr4								
2010	147004	13166	2400	-240				
2011	627628	65838	12540	-80995				
Grand Total	2399100	200000	74215	-80995				

The highest total number of stocks purchased or sold was in the 4^{th} Quarter of 2011 and minimum in the 4^{th} Quarter of 2010. Maximum of stocks sold, returned and the total transaction of the stocks was in the 4^{th} Quarter of 2011 and we can verify with the fact that the stocks market is on growth from comparing from 2010 and hence, the data is true without any manipulations.

1. Comparing Unit Price of stocks by Year and Quarter.

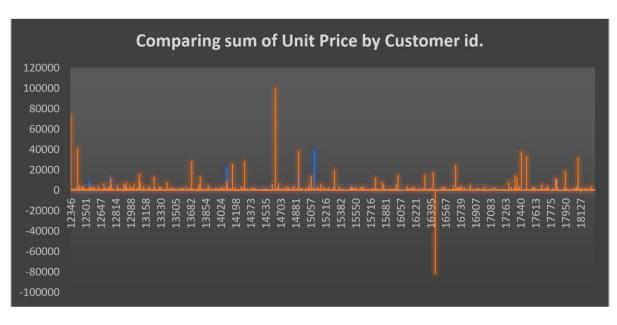


Row Labels	Sum of Unit Price	Min of Unit Price	Average of Unit Price	Max of Unit Price
<12/1/2010				
<12/1/2010				
Qtr1				
2011	115999.38	0	3.372565198	1715.85
Qtr2				
2011	175317.191	0	4.453970606	38970
Qtr3				
2011	144912.883	0	3.067653485	3155.95
Qtr4				
2010	42120.86	0	3.199214644	295
2011	212519.49	0	3.227915338	4161.06
Grand Total	690869.804	0	3.45434902	38970

Maximum unit price was in the 4^{th} Quarter of 2011 and minimum was in the 4^{th} Quarter of 2010. The average price was highest in the 2^{nd} Quarter of 2011 and it was minimum in the 4^{th} Quarter of 2010. The Maximum sum of unit price was in the 4^{th} Quarter of 2011.

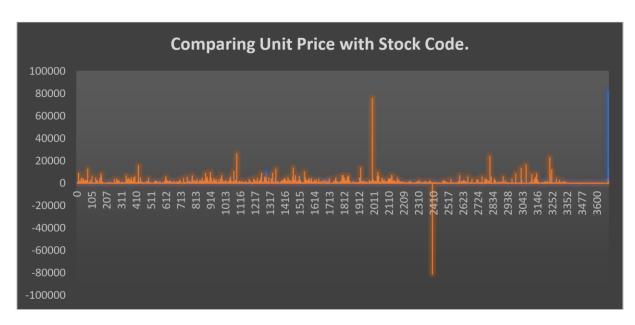
As, from the above tables we have seen that the maximum stocks quantity, maximum unit price maximum transaction was done in the 4th Quarter of 2011 that prove the fact that the stocks market are on demand and are increasing as more and more people are getting involve with the stock market.

Comparing Unit Price of stocks by Customer Id.



Customer with customer id 15098 has the maximum total unit price for its stocks and with total of 908 Stock Quantity.

Comparing Unit Price of stocks by Stock Code.



Stock code 3683 has the maximum total unit price for its stocks and has the stock quantity of 1510 stocks.

2. Predictive Modelling using R-software.

```
> #To import the training dataset.
> dataset = read.csv(file.choose())
> #To view the dataset
> View(dataset)
> #Removing the unwanted variable Invoice date and time that is not in the proper
> # format.
> dataset = dataset[-5]
```

•	InvoiceNo ‡	StockCode [‡]	Description ‡	Quantity ‡	UnitPrice ‡	CustomerID ‡	Country [‡]
1	6141	1583	144	3	3.75	14056	35
2	6349	1300	3682	6	1.95	13098	35
3	16783	2178	1939	4	5.95	15044	35
4	16971	2115	2983	1	0.83	15525	35
5	6080	1210	2886	12	1.65	13952	35
6	17388	495	3247	5	1.65	15351	35
7	18494	165	3377	1	1.25	12748	35
8	17109	2597	3435	1	1.25	16255	35
9	17143	1945	2352	1	5.75	17841	35
10	8422	3311	2502	6	2.95	13849	35
11	3548	321	2732	9	5.95	14466	35
12	4993	1236	1802	4	4.25	13015	35
13	3140	1508	3495	30	1.65	14646	23
14	1521	1417	815	12	1.49	13081	35
15	3621	3045	1635	1	1.69	16225	35
16	6186	1249	137	1	4.25	16393	35
17	18165	// 38	34.00	16	2.46	14096	35

```
> #Splitting the dataset into the Training set and Test set
> # # install.packages('caTools')
> library(caTools)
Warning message:
package 'caTools' was built under R version 3.6.3
> set.seed(123)
> split = sample.split(dataset$UnitPrice, SplitRatio = 2/3)
> training_set = subset(dataset, split == TRUE)
> test_set = subset(dataset, split == FALSE)
> #To check the correlation between the variables.
> cor(dataset)
                                        Description
               InvoiceNo
                            StockCode
                                                          Quantity
                                                                        UnitPrice
InvoiceNo
             1.000000000 0.086179070 0.0269460886 -0.0116752697 0.0069203647
StockCode 0.086179070 1.000000000 -0.0109126221 -0.0018360450 0.0175500942 Description 0.026946089 -0.010912622 1.0000000000 -0.0006121426 -0.0004513494
Quantity -0.011675270 -0.001836045 -0.0006121426 1.0000000000 -0.0009658447
UnitPrice 0.006920365 0.017550094 -0.0004513494 -0.0009658447 1.0000000000 
CustomerID -0.008351046 0.002970887 -0.0041180254 -0.0070690183 -0.0042361696
Country
             CustomerID
                               Country
            -0.008351046 0.003586286
InvoiceNo
StockCode
            0.002970887 0.008025297
Description -0.004118025 -0.014692252
          -0.007069018 -0.009544674
Quantity
UnitPrice
            -0.004236170 -0.004262872
CustomerID
             1.000000000
                          0.389674239
             0.389674239 1.000000000
Country
```

From the above sniped we can check the correlation between various variables. We can see that the there is weak positive relation between Invoice no, Stock code, and unit price. There is weak negative relation between Description, Quantity, Customer id, Country with Unit price.

3. Fitting Multiple Linear Regression Model using stepwise regression method in both directions.

```
> #Fitting of the linear regression model
> null = lm(UnitPrice~1, data = training_set)
> full = lm(UnitPrice~., data = training_set)
> model = step(null, scope = list(upper=full), data = data1, direction = "both")
Start: AIC=1251824
UnitPrice ~ 1
Df Sum of Sq RSS AIC
+ StockCode 1 492833 1588253455 1251785
+ InvoiceNo 1 89895 1588656393 1251818
Step: AIC=1251785
UnitPrice ~ StockCode
             Df Sum of Sq
                                  RSS
            1 58199 1588195256 1251782
+ InvoiceNo
```

```
Step: AIC=1251782
UnitPrice ~ StockCode + InvoiceNo
             Df Sum of Sq
                                 RSS
> summary(model)
lm(formula = UnitPrice ~ StockCode + InvoiceNo, data = training_set)
Residuals:
  Min 1Q Median 3Q Max
-9 -3 -1 1 38961
Coefficients:
             Estimate Std. Error t value Pr(>|t|)
(Intercept) -1.077e+00 8.001e-01 -1.346 0.1784
StockCode 2.211e-03 3.552e-04 6.223 4.89e-10 ***
InvoiceNo 1.194e-04 5.399e-05 2.211 0.0271 *
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 109.1 on 133379 degrees of freedom
Multiple R-squared: 0.0003468, Adjusted R-squared:
F-statistic: 23.14 on 2 and 133379 DF, p-value: 8.973e-11
```

Hence, the best multiple Linear Regression Model for predicting unit price of stock is **Unit Price** ~ **Stock code** + **Invoice No** with adjusted R-square as 0.033 % that is too low and hence multiple linear regression model is not working well with this model.

4. Prediction of Unit Price with Multiple Linear Regression Model.

```
> y = test_set$UnitPrice
> test_set1 = test_set[-5]
  y_pred = predict(model,test_set1)
> head(y_pred)
                           12
5.7408544 1.4953056 2.2514149 5.2546998 2.3374006 0.8993805
> residuals = sum((y-y_pred)^2)/length(y)
 residuals
[1] 514.1727
```

The mean error sum of squares for multiple linear regression model is **514.1727.**

5. Fitting Decision Tree Regression Model.

```
> # Fitting Decision Tree Regression to the dataset
> # install.packages('rpart')
> library(rpart)
Warning message:
package 'rpart' was built under R version 3.6.3
> regressor = rpart(formula = UnitPrice ~ .,
                    data = training_set,
                    control = rpart.control(minsplit = 1))
> # Predicting a new result with Decision Tree Regression
> y_pred1 = predict(regressor, test_set1)
> error = sum((y-y_pred1)^2)/length(y)
> error
[1] 513.6874
```

With decision tree regression model the mean error sum of squares is **513.6874** that is less than the multiple linear regression model. Hence, we will prefer Decision Tree Regression model over multiple linear regression model.

Checking the predicted values of unit price for test data set.

```
> head(y_pred1)
[1] 3.5936 3.5936 3.5936 3.5936 3.5936
```

• Observations:

- 1. Stock code 3683 has the maximum total unit price for its stocks and has the stock quantity of 1510 stocks.
- 2. Customer with customer id 15098 has the maximum total unit price for its stocks and with total of 908 Stock Quantity.
- 3. Maximum unit price was in the 4th Quarter of 2011 and minimum was in the 4th Quarter of 2010. The average price was highest in the 2nd Quarter of 2011 and it was minimum in the 4th Quarter of 2010. The Maximum sum of unit price was in the 4th Quarter of 2011.
- 4. There is weak positive relation between Invoice no, Stock code, and unit price. There is weak negative relation between Description, Quantity, Customer id, Country with Unit price.

• Conclusion:

1. Decision Tree regression works well than the multiple linear regression model for predicting Unit Price.