

Code ▼

Exploratory Data Analysis

Exploratory Data Analysis on Multivariate Dataset.

Steps Involved:

1. Loading and Reading the dataset
2. Insights about dataset
 - a. Structure
 - b. Dimensions
 - c. Data types of predicting variables
 - d. Summary of the dataset
 - e. Removing duplicate columns such as Region
3. Data Cleansing
 - a. Solved mapping issues between variable and its corresponding indicators using Excel.
 - b. Converted numeric variables with NA's to 0. c)Computed the summary of the new dataset.
4. Data Visualization
 - a. Plotted Box Plots and Strip charts to understand the data distribution and to detect outliers. Skewness in data and outliers were observed.
 - b. Plotted Scatter Plot Matrix using GGally library to understand the correlation between other variables as well as CPC.

–Loading and reading the dataset

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```
Data<-read.csv('D:/Rutgers Study Material/MultivariateData1.csv')
# top 5 columns of the dataset
head(Data)
```

ï..Region <fctr>	Region_Indicators <int>	City <fctr>	City_indicators <int>	SupplyVendor <fctr>	SupplyVendors_Indi
1 Hawaii	60	'Aiea	2208	beanstock	
2 Hawaii	60	'Aiea	2208	brightroll	
3 Hawaii	60	'Aiea	2208	brightroll	
4 Hawaii	60	'Aiea	2208	brightroll	
5 Hawaii	60	'Aiea	2208	brightroll	
6 Hawaii	60	'Aiea	2208	brightroll	

6 rows | 1-7 of 23 columns

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```
#Names of the columns in dataset  
names(Data)
```

```
[1] "i..Region"      "Region_Indicators"  
[3] "City"           "City_indicators"  
[5] "SupplyVendor"   "SupplyVendors_Indicators"  
[7] "OS"             "OS_Indicators"  
[9] "Browser"        "Browser_Indicators"  
[11] "DeviceType"     "DeviceType_Indicators"  
[13] "Impression_Day" "Impression_Time"  
[15] "Impressions"    "Clicks"  
[17] "CTR"            "CPC"  
[19] "VCR"            "CPV"  
[21] "Completes"      "Total_Spend"  
[23] "CPCV"
```

There are 23 columns present.

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```
#Structure of the data  
str(Data)
```

```
'data.frame':  472666 obs. of  23 variables:
 $ i..Region      : Factor w/ 27 levels "Alabama","Alaska",...: 13 13 13 13 13 13 13 13 13 13 ...
 $ Region_Indicators : int  60 60 60 60 60 60 60 60 60 60 ...
 $ City           : Factor w/ 5146 levels "'Aiea","'Ewa Beach",...: 1 1 1 1 1 1 1 1 1 1 ...
 ...
 $ City_indicators  : int  2208 2208 2208 2208 2208 2208 2208 2208 2208 2208 ...
 $ SupplyVendor     : Factor w/ 23 levels "adaptv","adconductor",...: 5 7 7 7 7 7 7 7 7 7 ...
 ...
 $ SupplyVendors_Indicators: int  3 4 4 4 4 4 4 4 4 4 ...
 $ OS               : Factor w/ 26 levels "Android23","Android40",...: 21 3 4 4 6 7 7 7 7 7 ...
 $ OS_Indicators    : int  1 8 9 9 11 12 12 12 12 13 ...
 $ Browser          : Factor w/ 12 levels "Chrome","Edge",...: 11 1 1 1 1 1 1 1 1 1 ...
 $ Browser_Indicators : int  4 1 1 1 1 1 1 1 1 1 ...
 $ DeviceType       : Factor w/ 4 levels "Mobile","PC",...: 4 4 4 4 4 4 4 4 4 4 ...
 $ DeviceType_Indicators : int  1 1 1 1 1 1 1 1 1 1 ...
 $ Impression_Day   : int  1 3 5 5 4 4 5 5 5 4 ...
 $ Impression_Time  : Factor w/ 84836 levels "0:00:04","0:00:07",...: 83360 3857 75392 75506 39916 73326 8081 8637 14189 74577 ...
 $ Impressions      : int  1 1 1 1 1 1 1 1 1 1 ...
 $ Clicks           : int  0 0 0 0 0 0 0 0 0 0 ...
 $ CTR              : num  0 0 0 0 0 0 0 0 0 0 ...
 $ CPC              : num  NA NA NA NA NA NA NA NA NA NA ...
 $ VCR              : num  0 1 1 0 1 1 1 0 0 1 ...
 $ CPV              : num  0.01271 0.00997 0.01128 0.00941 0.01025 ...
 $ Completes        : int  0 1 1 0 1 1 1 0 0 1 ...
 $ Total_Spend      : num  0.01271 0.00997 0.01128 0.00941 0.01025 ...
 $ CPCV             : num  NA 0.00997 0.01128 NA 0.01025 ...
```

There are 7 factors or categorical variables. Lots of NA's or missing values were oobserved.

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```
# Dimension of the data
dim(Data)
```

```
[1] 472666    23
```

Hide

```
# Removing the extra columns
drops <- c("i..Region","City","SupplyVendor","OS","Browser","DeviceType","Impression_Time")
New_data<-Data[ , !(names(Data) %in% drops)]
dim(New_data)
```

```
[1] 472666    16
```

Since we have column names as well as their indicators, it's always better to remove redundant information.

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```
# Checking the datatype of each column  
attach(New_data)
```

The following objects are masked from New_data (pos = 3):

Browser_Indicators, City_indicators, Clicks,
Completes, CPC, CPCV, CPV, CTR, DeviceType_Indicators,
Impression_Day, Impressions, OS_Indicators,
Region_Indicators, SupplyVendors_Indicators,
Total_Spend, VCR

The following objects are masked from New_data (pos = 4):

Browser_Indicators, City_indicators, Clicks,
Completes, CPC, CPCV, CPV, CTR, DeviceType_Indicators,
Impression_Day, Impressions, OS_Indicators,
Region_Indicators, SupplyVendors_Indicators,
Total_Spend, VCR

The following objects are masked from New_data (pos = 5):

Browser_Indicators, City_indicators, Clicks,
Completes, CPC, CPCV, CPV, CTR, DeviceType_Indicators,
Impression_Day, Impressions, OS_Indicators,
Region_Indicators, SupplyVendors_Indicators,
Total_Spend, VCR

The following objects are masked from New_data (pos = 6):

Browser_Indicators, City_indicators, Clicks,
Completes, CPC, CPCV, CPV, CTR, DeviceType_Indicators,
Impression_Day, Impressions, OS_Indicators,
Region_Indicators, SupplyVendors_Indicators,
Total_Spend, VCR

The following objects are masked from New_data (pos = 7):

Browser_Indicators, City_indicators, Clicks,
Completes, CPC, CPCV, CPV, CTR, DeviceType_Indicators,
Impression_Day, Impressions, OS_Indicators,
Region_Indicators, SupplyVendors_Indicators,
Total_Spend, VCR

The following objects are masked from New_data (pos = 10):

Browser_Indicators, City_indicators, Clicks,
Completes, CPC, CPCV, CPV, CTR, DeviceType_Indicators,
Impression_Day, Impressions, OS_Indicators,
Region_Indicators, SupplyVendors_Indicators,
Total_Spend, VCR

The following objects are masked from New_data (pos = 11):

Browser_Indicators, City_indicators, Clicks,
Completes, CPC, CPCV, CPV, CTR, DeviceType_Indicators,
Impression_Day, Impressions, OS_Indicators,

```
Region_Indicators, SupplyVendors_Indicators,  
Total_Spend, VCR
```

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```
class(Region_Indicators)
```

```
[1] "integer"
```

Hide

```
class(City_indicators)
```

```
[1] "integer"
```

Hide

```
class(SupplyVendors_Indicators)
```

```
[1] "integer"
```

Hide

```
class(OS_Indicators)
```

```
[1] "integer"
```

Hide

```
class(Browser_Indicators)
```

```
[1] "integer"
```

Hide

```
class(DeviceType_Indicators)
```

```
[1] "integer"
```

Hide

```
class(Impression_Day)
```

```
[1] "integer"
```

Hide

```
class(Impressions)
```

```
[1] "integer"
```

[Hide](#)

```
class(Clicks)
```

```
[1] "integer"
```

[Hide](#)

```
class(CTR)
```

```
[1] "numeric"
```

[Hide](#)

```
class(CPC)
```

```
[1] "numeric"
```

[Hide](#)

```
class(VCR)
```

```
[1] "numeric"
```

[Hide](#)

```
class(CPV)
```

```
[1] "numeric"
```

[Hide](#)

```
class(Completes)
```

```
[1] "integer"
```

[Hide](#)

```
class(Total_Spend)
```

```
[1] "numeric"
```

Hide

```
class(CPCV)
```

```
[1] "numeric"
```

Hide

```
# Analyzing missing values
sapply(New_data,function(x) sum(is.na(x)))
```

Region_Indicators	City_indicators
0	0
SupplyVendors_Indicators	OS_Indicators
0	0
Browser_Indicators	DeviceType_Indicators
0	0
Impression_Day	Impressions
0	0
Clicks	CTR
0	0
CPC	VCR
469436	0
CPV	Completes
13443	0
Total_Spend	CPCV
0	190689

A lot of indicators were missing from the data. On analyzing the file "VLookUP" was not working properly. Steps Taken: Mapped the indicators using excel.

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```
summary(New_data)
```


Region_Indicators City_indicators SupplyVendors_Indicators

Min. : 2.00	Min. : 1	Min. : 1.000
1st Qu.: 3.00	1st Qu.: 946	1st Qu.: 4.000
Median :31.00	Median :1672	Median : 4.000
Mean :24.09	Mean :1895	Mean : 8.146
3rd Qu.:39.00	3rd Qu.:2590	3rd Qu.:15.000
Max. :60.00	Max. :5151	Max. :23.000

OS_Indicators Browser_Indicators DeviceType_Indicators

Min. : 1.000	Min. : 1.000	Min. :1.000
1st Qu.: 1.000	1st Qu.: 1.000	1st Qu.:1.000
Median : 3.000	Median : 1.000	Median :2.000
Mean : 4.703	Mean : 2.801	Mean :1.883
3rd Qu.: 5.000	3rd Qu.: 4.000	3rd Qu.:2.000
Max. :25.000	Max. :12.000	Max. :4.000

Impression_Day Impressions Clicks

Min. :1.000	Min. : 1.000	Min. :0.000000
1st Qu.:2.000	1st Qu.: 1.000	1st Qu.:0.000000
Median :4.000	Median : 1.000	Median :0.000000
Mean :4.081	Mean : 1.055	Mean :0.006846
3rd Qu.:6.000	3rd Qu.: 1.000	3rd Qu.:0.000000
Max. :7.000	Max. :120.000	Max. :2.000000

CTR

CPC

VCR

Min. :0.00000	Min. :0	Min. :0.0000
1st Qu.:0.00000	1st Qu.:0	1st Qu.:0.0000
Median :0.00000	Median :0	Median :1.0000
Mean :0.00645	Mean :0	Mean :0.5925
3rd Qu.:0.00000	3rd Qu.:0	3rd Qu.:1.0000
Max. :1.00000	Max. :0	Max. :1.0000

NA's :469436

CPV

Completes

Total_Spend

Min. :0.000	Min. : 0.0000	Min. :0.000000
1st Qu.:0.010	1st Qu.: 0.0000	1st Qu.:0.009459
Median :0.011	Median : 1.0000	Median :0.011098
Mean :0.012	Mean : 0.6258	Mean :0.011581
3rd Qu.:0.014	3rd Qu.: 1.0000	3rd Qu.:0.013244
Max. :0.038	Max. :120.0000	Max. :0.092112

NA's :13443

CPCV

Min. :0.00
1st Qu.:0.01
Median :0.01
Mean :0.01
3rd Qu.:0.01
Max. :0.05
NA's :190689

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```
New_data$CPV[ is.na(New_data$CPV)] <- 0
New_data$CPC[ is.na(New_data$CPC)] <- 0
New_data$CPCV[ is.na(New_data$CPCV)] <- 0
```

Hide

```
summary(New_data)
```

```
Region_Indicators City_indicators SupplyVendors_Indicators
Min.   : 2.00      Min.   : 1      Min.   : 1.000
1st Qu.: 3.00      1st Qu.: 946      1st Qu.: 4.000
Median :31.00      Median :1672      Median : 4.000
Mean   :24.09      Mean   :1895      Mean   : 8.146
3rd Qu.:39.00      3rd Qu.:2590      3rd Qu.:15.000
Max.   :60.00      Max.   :5151      Max.   :23.000
OS_Indicators  Browser_Indicators DeviceType_Indicators
Min.   : 1.000    Min.   : 1.000      Min.   :1.000
1st Qu.: 1.000    1st Qu.: 1.000      1st Qu.:1.000
Median : 3.000    Median : 1.000      Median :2.000
Mean   : 4.703    Mean   : 2.801      Mean   :1.883
3rd Qu.: 5.000    3rd Qu.: 4.000      3rd Qu.:2.000
Max.   :25.000    Max.   :12.000      Max.   :4.000
Impression_Day Impressions      Clicks
Min.   :1.000    Min.   : 1.000      Min.   :0.000000
1st Qu.:2.000    1st Qu.: 1.000      1st Qu.:0.000000
Median :4.000    Median : 1.000      Median :0.000000
Mean   :4.081    Mean   : 1.055      Mean   :0.006846
3rd Qu.:6.000    3rd Qu.: 1.000      3rd Qu.:0.000000
Max.   :7.000    Max.   :120.000      Max.   :2.000000
      CTR              CPC              VCR
Min.   :0.00000      Min.   :0.000e+00      Min.   :0.0000
1st Qu.:0.00000      1st Qu.:0.000e+00      1st Qu.:0.0000
Median :0.00000      Median :0.000e+00      Median :1.0000
Mean   :0.00645      Mean   :7.941e-05      Mean   :0.5925
3rd Qu.:0.00000      3rd Qu.:0.000e+00      3rd Qu.:1.0000
Max.   :1.00000      Max.   :4.450e-02      Max.   :1.0000
      CPV              Completes      Total_Spend
Min.   :0.000000      Min.   : 0.0000      Min.   :0.000000
1st Qu.:0.009381      1st Qu.: 0.0000      1st Qu.:0.009459
Median :0.011073      Median : 1.0000      Median :0.011098
Mean   :0.011272      Mean   : 0.6258      Mean   :0.011581
3rd Qu.:0.013164      3rd Qu.: 1.0000      3rd Qu.:0.013244
Max.   :0.038316      Max.   :120.0000      Max.   :0.092112
      CPCV
Min.   :0.000000
1st Qu.:0.000000
Median :0.009233
Mean   :0.007089
3rd Qu.:0.012036
Max.   :0.049803
```

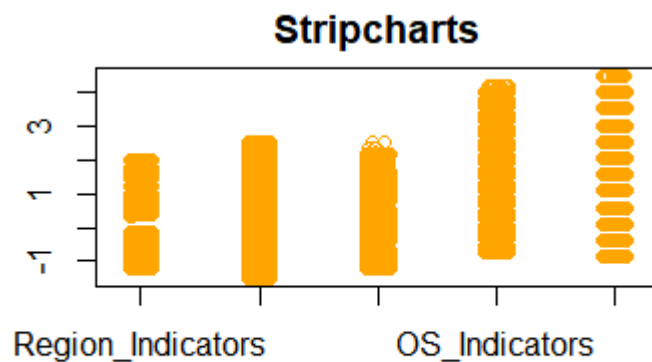
We can observe skewness in data as mean is either greater than or less than median.

Visualizing the Data

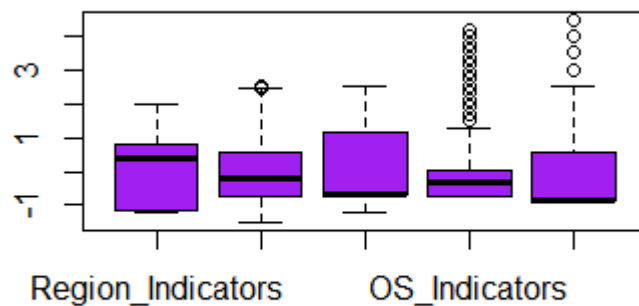
Plotting stripcharts and boxplots side-by-side can be useful to visualize the spread and distribution of data as well as analyzing outliers.

[Hide](#)

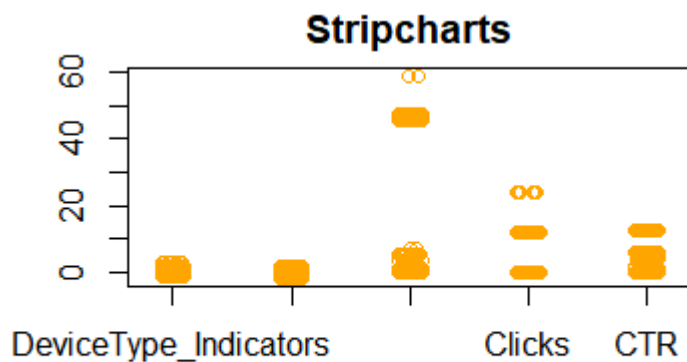
```
#New_data$City_indicators<-as.numeric(levels(New_data$City_indicators))[New_data$City_indicators]
## Stripcharts
numeric_data <- New_data[,c(1:5)]
numeric_data <- data.frame(scale(numeric_data ))
strip<-stripchart(numeric_data,
  vertical = TRUE,
  method = "jitter",
  col = "orange",
  pch=1,
  main="Stripcharts")
```


[Hide](#)

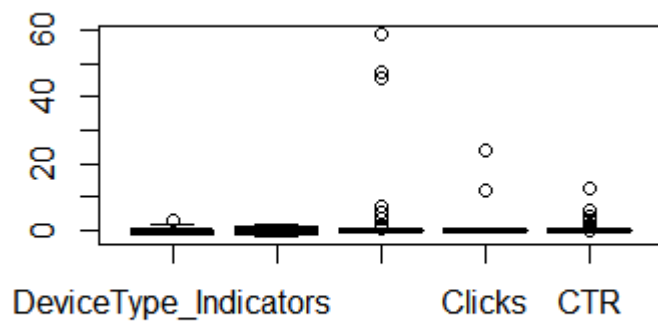
```
box<-boxplot(numeric_data,col='Purple')
```


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```
## Stripcharts
numeric_data <- New_data[,c(6:10)]
numeric_data <- data.frame(scale(numeric_data ))
strip<-stripchart(numeric_data,
  vertical = TRUE,
  method = "jitter",
  col = "orange",
  pch=1,
  main="Stripcharts")
```

[Hide](#)

```
box<-boxplot(numeric_data)
```

[Hide](#)

```
## Stripcharts
numeric_data <- new[,c(10:16)]
numeric_data <- data.frame(scale(numeric_data ))
strip<-stripchart(numeric_data,
                  vertical = TRUE,
                  method = "jitter",
                  col = "orange",
                  pch=1,
                  main="Stripcharts")
box<-boxplot(numeric_data)
```

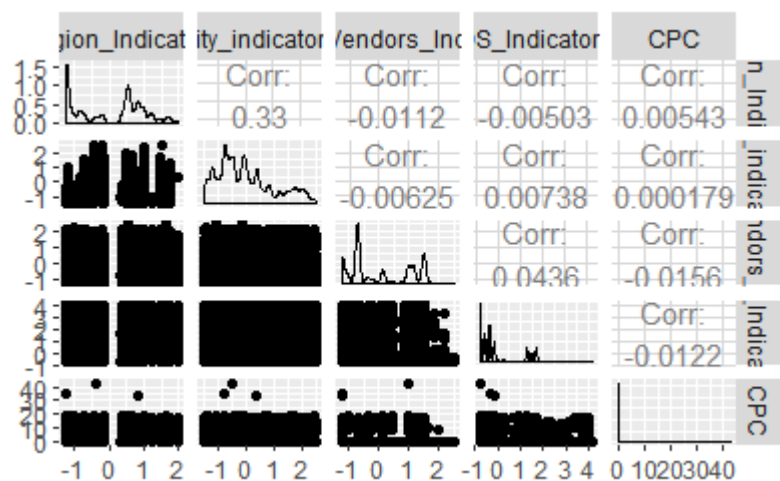
From the above plots we can confirm about skewness and presence of outliers as well.

Scatter Plot matrix is another important way to visualize data, its distribution and correlation with other variables.

[Hide](#)

```
numeric_data <- New_data[,c(1,2,3,4,11)]
numeric_data <- data.frame(scale(numeric_data ))
library("GGally")
ggpairs(numeric_data)
```

```
plot: [1,1] [=-----] 4% est: 0s
plot: [1,2] [===-----] 8% est: 4s
plot: [1,3] [====-----] 12% est: 5s
plot: [1,4] [=====-----] 16% est: 4s
plot: [1,5] [=====-----] 20% est: 3s
plot: [2,1] [=====-----] 24% est: 3s
plot: [2,2] [=====-----] 28% est: 6s
plot: [2,3] [=====-----] 32% est: 5s
plot: [2,4] [=====-----] 36% est: 5s
plot: [2,5] [=====-----] 40% est: 4s
plot: [3,1] [=====-----] 44% est: 4s
plot: [3,2] [=====-----] 48% est: 5s
plot: [3,3] [=====-----] 52% est: 5s
plot: [3,4] [=====-----] 56% est: 5s
plot: [3,5] [=====-----] 60% est: 4s
plot: [4,1] [=====-----] 64% est: 3s
plot: [4,2] [=====-----] 68% est: 4s
plot: [4,3] [=====-----] 72% est: 3s
plot: [4,4] [=====-----] 76% est: 3s
plot: [4,5] [=====-----] 80% est: 3s
plot: [5,1] [=====-----] 84% est: 2s
plot: [5,2] [=====-----] 88% est: 2s
plot: [5,3] [=====-----] 92% est: 1s
plot: [5,4] [=====-----] 96% est: 1s
plot: [5,5] [=====] 100% est: 0s
```



CPC, Region Indicators, City Indicators are positively correlated while CPC, Vendor Indicator and OS Indicator are negatively correlated.

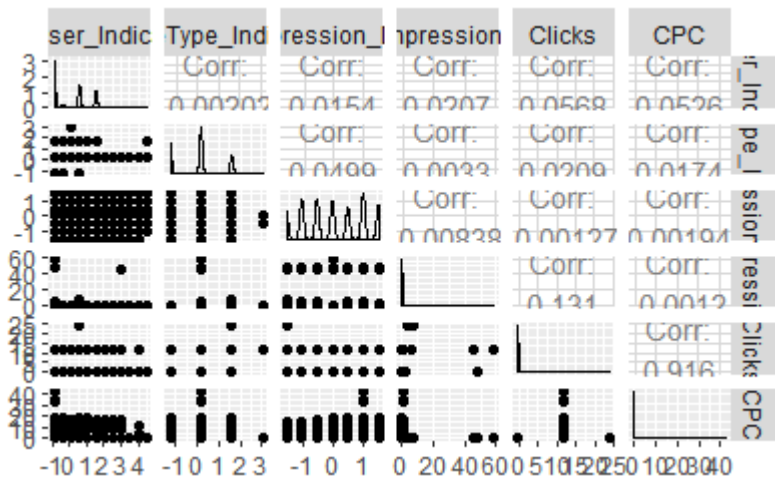
[Hide](#)

```
numeric_data <- New_data[,c(5,6,7,8,9,11)]
numeric_data <- data.frame(scale(numeric_data ))
library("GGally")
ggpairs(numeric_data)
```

```

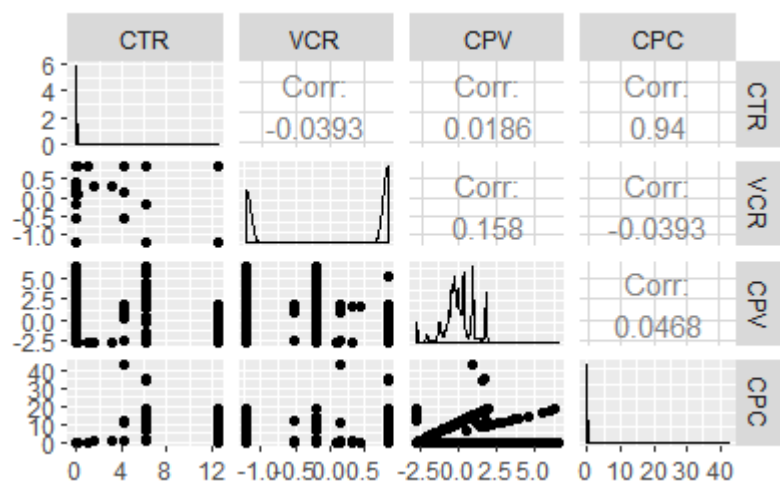
plot: [1,1] [=-----] 3% est: 0s
plot: [1,2] [==-----] 6% est: 5s
plot: [1,3] [===-----] 8% est: 4s
plot: [1,4] [====-----] 11% est: 4s
plot: [1,5] [=====-----] 14% est: 4s
plot: [1,6] [=====-----] 17% est: 4s
plot: [2,1] [=====-----] 19% est: 4s
plot: [2,2] [=====-----] 22% est: 7s
plot: [2,3] [=====-----] 25% est: 7s
plot: [2,4] [=====-----] 28% est: 7s
plot: [2,5] [=====-----] 31% est: 6s
plot: [2,6] [=====-----] 33% est: 6s
plot: [3,1] [=====-----] 36% est: 5s
plot: [3,2] [=====-----] 39% est: 7s
plot: [3,3] [=====-----] 42% est: 8s
plot: [3,4] [=====-----] 44% est: 7s
plot: [3,5] [=====-----] 47% est: 7s
plot: [3,6] [=====-----] 50% est: 6s
plot: [4,1] [=====-----] 53% est: 6s
plot: [4,2] [=====-----] 56% est: 6s
plot: [4,3] [=====-----] 58% est: 6s
plot: [4,4] [=====-----] 61% est: 6s
plot: [4,5] [=====-----] 64% est: 6s
plot: [4,6] [=====-----] 67% est: 5s
plot: [5,1] [=====-----] 69% est: 5s
plot: [5,2] [=====-----] 72% est: 5s
plot: [5,3] [=====-----] 75% est: 5s
plot: [5,4] [=====-----] 78% est: 4s
plot: [5,5] [=====-----] 81% est: 4s
plot: [5,6] [=====-----] 83% est: 3s
plot: [6,1] [=====-----] 86% est: 3s
plot: [6,2] [=====-----] 89% est: 2s
plot: [6,3] [=====-----] 92% est: 2s
plot: [6,4] [=====-----] 94% est: 1s
plot: [6,5] [=====-----] 97% est: 1s
plot: [6,6] [=====] 100% est: 0s

```



```
numeric_data <- New_data[,c(10,12,13,11)]
numeric_data <- data.frame(scale(numeric_data ))
library("GGally")
ggpairs(numeric_data)
```

```
plot: [1,1] [=====] 6% est: 0s
plot: [1,2] [=====] 12% est: 2s
plot: [1,3] [=====] 19% est: 1s
plot: [1,4] [=====] 25% est: 1s
plot: [2,1] [=====] 31% est: 1s
plot: [2,2] [=====] 38% est: 3s
plot: [2,3] [=====] 44% est: 3s
plot: [2,4] [=====] 50% est: 2s
plot: [3,1] [=====] 56% est: 2s
plot: [3,2] [=====] 62% est: 2s
plot: [3,3] [=====] 69% est: 2s
plot: [3,4] [=====] 75% est: 2s
plot: [4,1] [=====] 81% est: 1s
plot: [4,2] [=====] 88% est: 1s
plot: [4,3] [=====] 94% est: 1s
plot: [4,4] [=====] 100% est: 0s
```



CTR,CPC,CPV are positively correlated and VCR negatively.

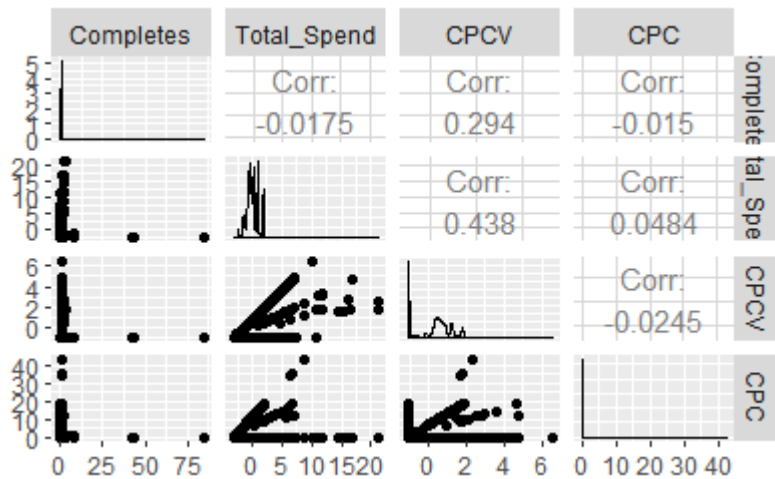
```
numeric_data <- New_data[,c(14,15,16,11)]
numeric_data <- data.frame(scale(numeric_data ))
library("GGally")
ggpairs(numeric_data)
```



```

plot: [1,1] [=====] 6% est: 0s
plot: [1,2] [=====] 12% est: 3s
plot: [1,3] [=====] 19% est: 2s
plot: [1,4] [=====] 25% est: 2s
plot: [2,1] [=====] 31% est: 2s
plot: [2,2] [=====] 38% est: 3s
plot: [2,3] [=====] 44% est: 3s
plot: [2,4] [=====] 50% est: 2s
plot: [3,1] [=====] 56% est: 2s
plot: [3,2] [=====] 62% est: 2s
plot: [3,3] [=====] 69% est: 2s
plot: [3,4] [=====] 75% est: 2s
plot: [4,1] [=====] 81% est: 1s
plot: [4,2] [=====] 88% est: 1s
plot: [4,3] [=====] 94% est: 1s
plot: [4,4] [=====] 100% est: 0s

```



CPC, Completes, CPCV are negatively correlated and total spend is positively correlated.