

CHAMPO CARPETS: **PERFORMING EDA ON DATASET** (DERIVING INSIGHTS AND CLEANING)

```
CC <- read_excel("C:/Users/Bodake/Desktop/UIC Sem 2/Data Mining IDS-
572/R/champo.xlsx")
View(CC)

mydata <- CC
mydata <- drop_na(CC)
str(mydata)

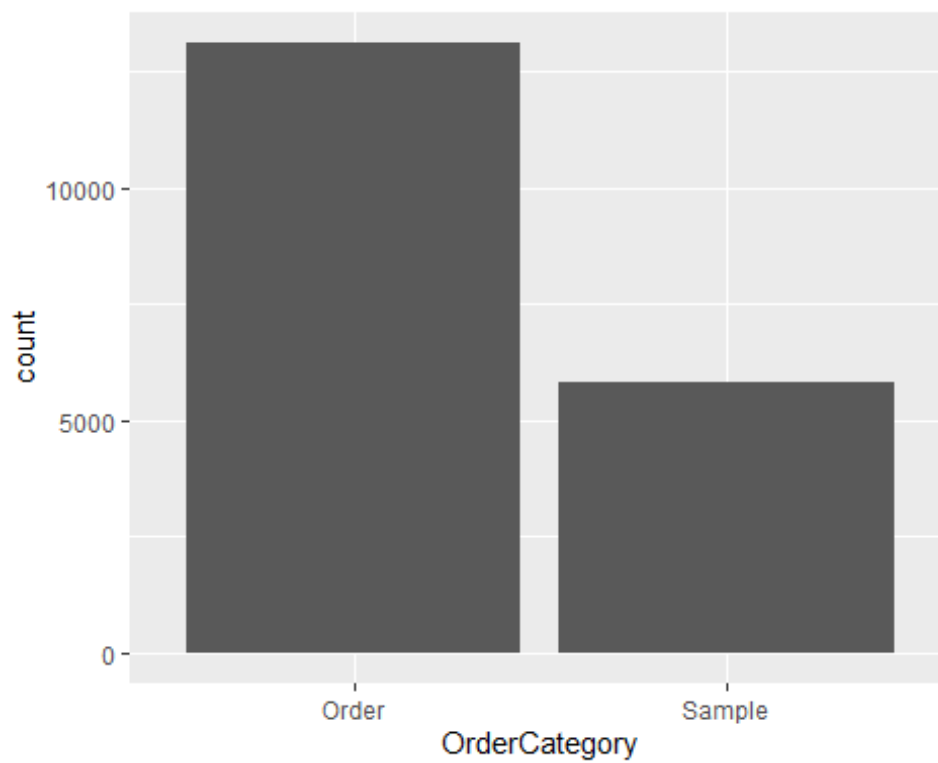
## tibble [18,946 x 16] (S3: tbl_df/tbl/data.frame)
## $ OrderType      : chr [1:18946] "Area Wise" "Area Wise" "Area Wise"
## "Area Wise" ...
## $ OrderCategory  : chr [1:18946] "Order" "Order" "Order" "Order" ...
## $ CustomerCode   : chr [1:18946] "H-1" "H-1" "H-1" "H-1" ...
## $ CountryName    : chr [1:18946] "USA" "USA" "USA" "USA" ...
## $ CustomerOrderNo: chr [1:18946] "1873354" "1873354" "1873354" "1918436"
## ...
## $ Custorderdate  : POSIXct[1:18946], format: "2017-01-16" "2017-01-16"
## ...
## $ UnitName       : chr [1:18946] "Ft" "Ft" "Ft" "Ft" ...
## $ QtyRequired    : num [1:18946] 2 2 2 5 5 4 6 16 2 4 ...
## $ TotalArea      : num [1:18946] 6 9 54 54 71.2 ...
## $ Amount         : num [1:18946] 12 18 108 270 356 ...
## $ ITEM_NAME      : chr [1:18946] "HAND TUFTED" "HAND TUFTED" "HAND
## TUFTED" "HAND TUFTED" ...
## $ QualityName    : chr [1:18946] "TUFTED 30C HARD TWIST" "TUFTED 30C HARD
## TWIST" "TUFTED 30C HARD TWIST" "TUFTED 30C HARD TWIST" ...
## $ DesignName     : chr [1:18946] "OLD LONDON [3715]" "OLD LONDON [3715]"
## "OLD LONDON [3715]" "OLD LONDON [3715]" ...
## $ ColorName      : chr [1:18946] "BEIGE" "BEIGE" "BEIGE" "BEIGE" ...
## $ ShapeName      : chr [1:18946] "REC" "REC" "REC" "REC" ...
## $ AreaFt         : num [1:18946] 6 9 54 54 71.2 ...

count(mydata)

## # A tibble: 1 x 1
##       n
##   <int>
## 1 18946

ggplot(data=mydata) + geom_bar(mapping=aes(x=OrderCategory))
```

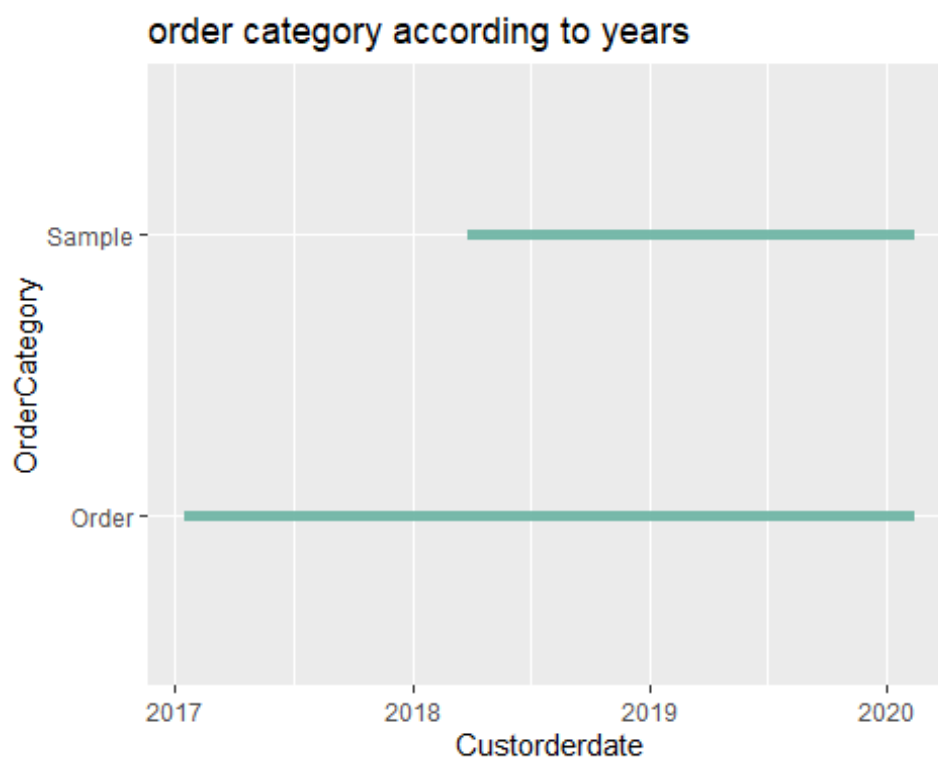
INSIGHTS FOR CHAMPO RAW DATASET



Here, number of 'Orders' in 'Order Category' is higher than 'Sample' category.

INSIGHTS FOR CHAMPO RAW DATASET

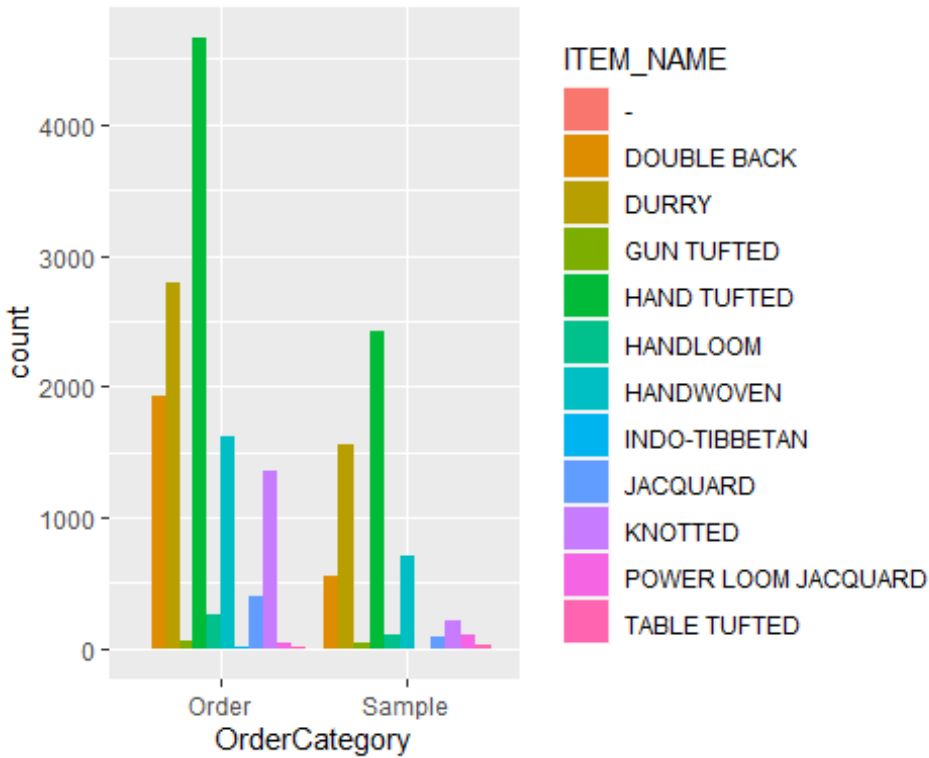
```
ggplot(mydata, aes(x=Custorderdate, y=OrderCategory)) +  
  geom_line( color="#69b3a2", size=2, alpha=0.9, linetype=1.5) +  
  ggtitle("order category according to years")
```



Here, there were no samples sent out in year 2017

INSIGHTS FOR CHAMPO RAW DATASET

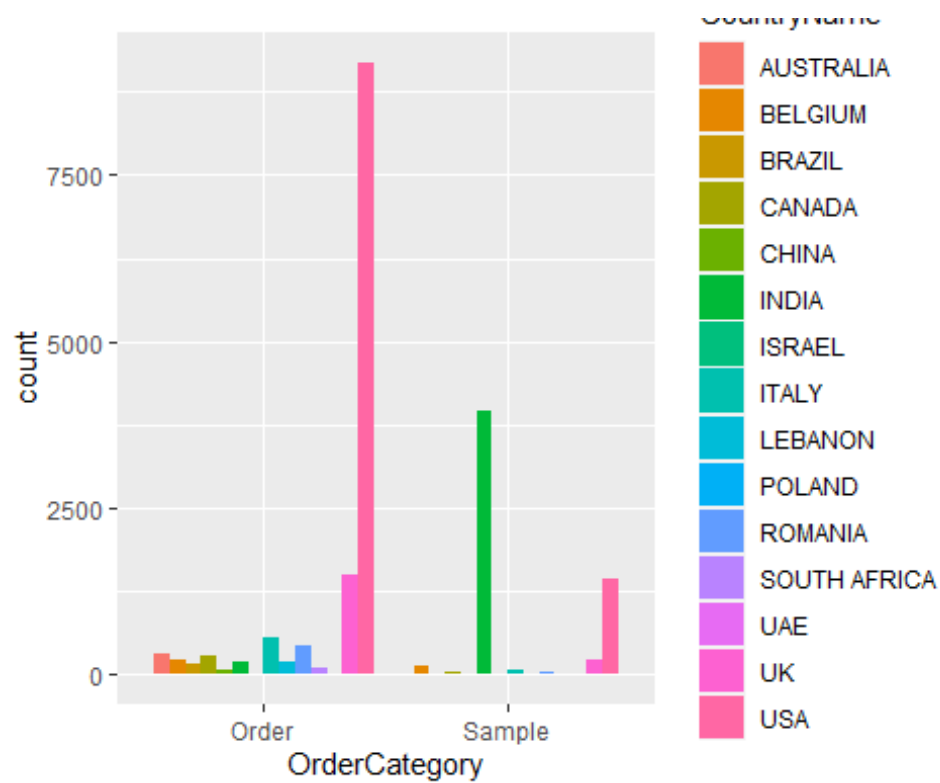
```
ggplot(mydata, aes(x = OrderCategory, fill =  
                    ITEM_NAME)) +  
  geom_bar(position = "dodge")
```



Here, as the “Hand-Tufted” carpets use base-material and entitled the least production effort, hence, generates maximum revenue for the organization in both order and sample category

INSIGHTS FOR CHAMPO RAW DATASET

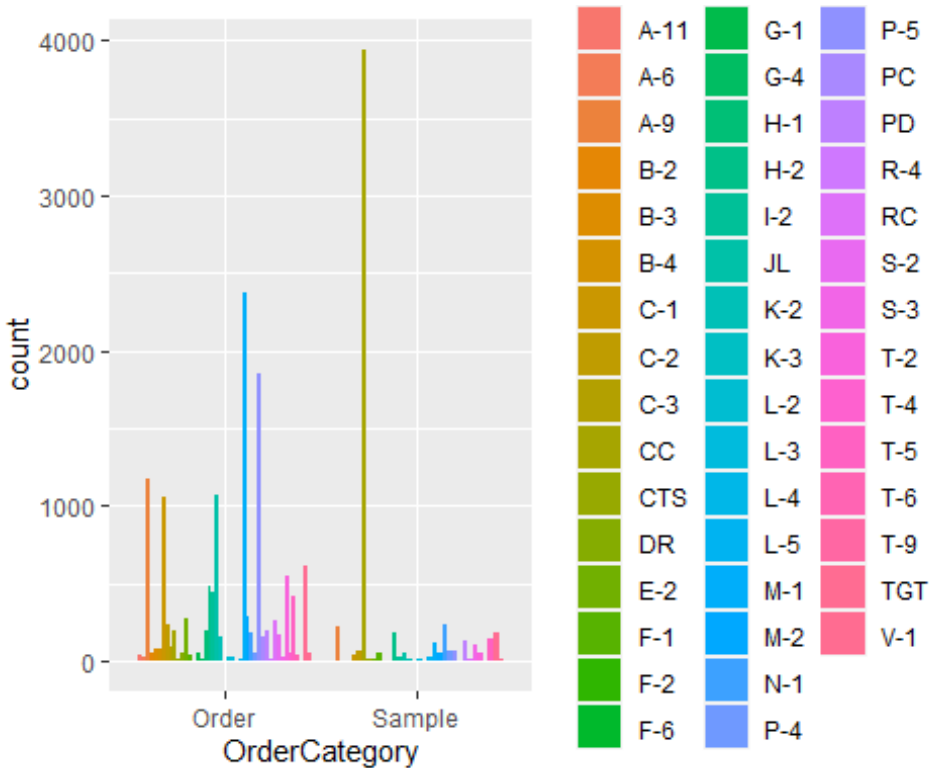
```
ggplot(mydata, aes(x = OrderCategory, fill =  
                    CountryName)) +  
  geom_bar(position = "dodge")
```



The maximum consumption of product was in USA followed by India and UK.
As, many Indians use Durries variants of kilim.

INSIGHTS FOR CHAMPO RAW DATASET

```
ggplot(mydata, aes(x = OrderCategory, fill =  
                  CustomerCode)) +  
  geom_bar(position = "dodge")
```



From the graph we can infer that more samples were sent out to the customers with customer code 'CC' and a greater number of orders we placed by customers with customer code between 'L-5' and 'N-1'.

These are some of the insights that could be derived from the Dataset.

--END OF INSIGHTS SECTION--

2) What kind of analytics and machine learning algorithms (e.g., classification, regression, clustering, recommender systems and etc) can be used by Champo Carpets to solve their problems, and in general for value creation? Justify your choices.

Answer:

- Champo carpets can use various analytics and machine learning algorithms to solve their problems such as **Decision Tree, Random Forest, Logistic Regression, K-means Clustering, Neural Networks etc.**
- All these algorithms/methods can be implemented on the Champo Dataset in this assignment.
- Then by calculating the accuracy of the different models, the model with the highest accuracy would be the best for Champo Carpets business.

CHAMPO CARPETS: DESCISION TREE MODEL

```
library(stats)
library(dplyr)

## Warning: package 'dplyr' was built under R version 4.1.2
##
## 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

library(readxl)

## Warning: package 'readxl' was built under R version 4.1.2

library(tidyr)
library(ggplot2)

## Warning: package 'ggplot2' was built under R version 4.1.2

library(ISLR)

## Warning: package 'ISLR' was built under R version 4.1.2

library(rpart)
library(rpart.plot)

## Warning: package 'rpart.plot' was built under R version 4.1.2

library(psych)

## Warning: package 'psych' was built under R version 4.1.2
##
## Attaching package: 'psych'

## The following objects are masked from 'package:ggplot2':
##
##   %+%, alpha

Sampledata <- read_excel("C:/Users/Bodake/Desktop/UIC Sem 2/Data Mining IDS-5
72/R/sample.xlsx")
View(Sampledata)
```


DESCISION TREE MODEL

```
mydata <- Sampledata
mydata

## # A tibble: 5,820 x 25
##   CustomerCode CountryName  USA    UK Italy Belgium Romania Australia In
dia
##   <chr>         <chr>      <dbl> <dbl> <dbl>  <dbl>  <dbl>      <dbl> <d
bl>
##  1 A-11        USA          1     0     0      0      0          0
##  2 A-11        USA          1     0     0      0      0          0
##  3 A-11        USA          1     0     0      0      0          0
##  4 A-11        USA          1     0     0      0      0          0
##  5 A-9         USA          1     0     0      0      0          0
##  6 A-9         USA          1     0     0      0      0          0
##  7 A-9         USA          1     0     0      0      0          0
##  8 A-9         USA          1     0     0      0      0          0
##  9 A-9         USA          1     0     0      0      0          0
## 10 A-9         USA          1     0     0      0      0          0
## # ... with 5,810 more rows, and 16 more variables: QtyRequired <dbl>,
## #   ITEMNAME <chr>, HandTufted <dbl>, Durry <dbl>, DoubleBack <dbl>,
## #   HandWoven <dbl>, Knotted <dbl>, Jacquard <dbl>, Handloom <dbl>,
## #   Other <dbl>, ShapeName <chr>, REC <dbl>, Round <dbl>, Square <dbl>,
## #   AreaFt <dbl>, OrderConversion <dbl>

summary(mydata)

## CustomerCode      CountryName      USA      UK
## Length:5820      Length:5820      Min.   :0.0000  Min.   :0.00000
## Class :character  Class :character  1st Qu.:0.0000  1st Qu.:0.00000
## Mode  :character  Mode  :character  Median :0.0000  Median :0.00000
##                                     Mean  :0.2491  Mean   :0.03512
##                                     3rd Qu.:0.0000  3rd Qu.:0.00000
##                                     Max.   :1.0000  Max.   :1.00000
##                                     NA's   :39      NA's   :39
##      Italy      Belgium      Romania      Australia
## Min.   :0.00000  Min.   :0.00000  Min.   :0.00000  Min.   :0.00000
## 1st Qu.:0.00000  1st Qu.:0.00000  1st Qu.:0.00000  1st Qu.:0.00000
## Median :0.00000  Median :0.00000  Median :0.00000  Median :0.00000
## Mean   :0.00778  Mean   :0.02283  Mean   :0.00346  Mean   :0.00173
## 3rd Qu.:0.00000  3rd Qu.:0.00000  3rd Qu.:0.00000  3rd Qu.:0.00000
```

DESCISION TREE MODEL

```
## Max. :1.00000 Max. :1.00000 Max. :1.00000 Max. :1.00000
## NA's :39 NA's :39 NA's :39 NA's :39
## India QtyRequired ITEMNAME HandTufted
## Min. :0.0000 Min. : 1.000 Length:5820 Min. :0.0000
## 1st Qu.:0.0000 1st Qu.: 1.000 Class :character 1st Qu.:0.0000
## Median :1.0000 Median : 1.000 Mode :character Median :0.0000
## Mean :0.6817 Mean : 1.975 Mean :0.4167
## 3rd Qu.:1.0000 3rd Qu.: 1.000 3rd Qu.:1.0000
## Max. :1.0000 Max. :200.000 Max. :1.0000
## NA's :39
## Durry DoubleBack HandWoven Knotted
## Min. :0.0000 Min. :0.00000 Min. :0.0000 Min. :0.00000
## 1st Qu.:0.0000 1st Qu.:0.00000 1st Qu.:0.0000 1st Qu.:0.00000
## Median :0.0000 Median :0.00000 Median :0.0000 Median :0.00000
## Mean :0.2686 Mean :0.09519 Mean :0.1211 Mean :0.03729
## 3rd Qu.:1.0000 3rd Qu.:0.00000 3rd Qu.:0.0000 3rd Qu.:0.00000
## Max. :1.0000 Max. :1.00000 Max. :1.0000 Max. :1.00000
##
## Jacquard Handloom Other ShapeName
## Min. :0.00000 Min. :0.0000 Min. :0.00000 Length:5820
## 1st Qu.:0.00000 1st Qu.:0.0000 1st Qu.:0.00000 Class :character
## Median :0.00000 Median :0.0000 Median :0.00000 Mode :character
## Mean :0.01443 Mean :0.0177 Mean :0.02904
## 3rd Qu.:0.00000 3rd Qu.:0.0000 3rd Qu.:0.00000
## Max. :1.00000 Max. :1.0000 Max. :1.00000
##
## REC Round Square AreaFt
## Min. :0.0000 Min. :0.000000 Min. :0.00000 Min. : 0.6667
## 1st Qu.:1.0000 1st Qu.:0.000000 1st Qu.:0.00000 1st Qu.: 6.0000
## Median :1.0000 Median :0.000000 Median :0.00000 Median : 11.0000
## Mean :0.9864 Mean :0.009794 Mean :0.00378 Mean : 21.5558
## 3rd Qu.:1.0000 3rd Qu.:0.000000 3rd Qu.:0.00000 3rd Qu.: 39.8125
## Max. :1.0000 Max. :1.000000 Max. :1.00000 Max. :480.0000
##
## OrderConversion
## Min. :0.0000
## 1st Qu.:0.0000
## Median :0.0000
## Mean :0.2009
## 3rd Qu.:0.0000
## Max. :1.0000
##
str(mydata)

## tibble [5,820 x 25] (S3: tbl_df/tbl/data.frame)
## $ CustomerCode : chr [1:5820] "A-11" "A-11" "A-11" "A-11" ...
## $ CountryName : chr [1:5820] "USA" "USA" "USA" "USA" ...
## $ USA : num [1:5820] 1 1 1 1 1 1 1 1 1 1 ...
## $ UK : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
```

DESCISION TREE MODEL

```
## $ Italy      : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Belgium   : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Romania   : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Australia : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ India     : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ QtyRequired : num [1:5820] 2 2 2 1 5 5 5 5 11 2 ...
## $ ITEMNAME   : chr [1:5820] "DURRY" "DOUBLE BACK" "JACQUARD" "JACQUAR
D" ...
## $ HandTufted : num [1:5820] 0 0 0 0 0 0 0 0 0 1 ...
## $ Durry      : num [1:5820] 1 0 0 0 0 0 0 0 0 0 ...
## $ DoubleBack : num [1:5820] 0 1 0 0 1 1 1 1 0 0 ...
## $ HandWoven  : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Knotted    : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Jacquard   : num [1:5820] 0 0 1 1 0 0 0 0 1 0 ...
## $ Handloom   : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Other      : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ ShapeName  : chr [1:5820] "REC" "REC" "REC" "REC" ...
## $ REC        : num [1:5820] 1 1 1 1 1 1 1 1 1 1 ...
## $ Round      : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Square     : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ AreaFt     : num [1:5820] 4 4 4 42.3 1 ...
## $ OrderConversion: num [1:5820] 0 0 0 1 0 0 0 0 0 0 ...
```

```
mydata <- na.omit(mydata)
mydata <- drop_na(Sampledata)
mydata
```

```
## # A tibble: 5,781 x 25
##   CustomerCode CountryName  USA    UK Italy Belgium Romania Australia In
dia
##   <chr>          <chr>    <dbl> <dbl> <dbl>  <dbl>  <dbl>  <dbl> <d
bl>
## 1 A-11          USA        1     0     0      0      0      0
## 2 A-11          USA        1     0     0      0      0      0
## 3 A-11          USA        1     0     0      0      0      0
## 4 A-11          USA        1     0     0      0      0      0
## 5 A-9           USA        1     0     0      0      0      0
## 6 A-9           USA        1     0     0      0      0      0
## 7 A-9           USA        1     0     0      0      0      0
## 8 A-9           USA        1     0     0      0      0      0
## 9 A-9           USA        1     0     0      0      0      0
```

DESCISION TREE MODEL

```
## 10 A-9          USA          1      0      0      0      0      0
0
## # ... with 5,771 more rows, and 16 more variables: QtyRequired <dbl>,
## #   ITEMNAME <chr>, HandTufted <dbl>, Durry <dbl>, DoubleBack <dbl>,
## #   HandWoven <dbl>, Knotted <dbl>, Jacquard <dbl>, Handloom <dbl>,
## #   Other <dbl>, ShapeName <chr>, REC <dbl>, Round <dbl>, Square <dbl>,
## #   AreaFt <dbl>, OrderConversion <dbl>

data <- mydata[-1]
```

Training and testing data in 85% and 15% using gini index and info gain

```
set.seed(134)
indx <- sample(2, nrow(data), replace = TRUE, prob = c(0.7, 0.3))
train <- data[indx == 1, ]
nrow(train)

## [1] 4075

test <- data[indx == 2, ]
nrow(test)

## [1] 1706

nrow(train)/nrow(data)

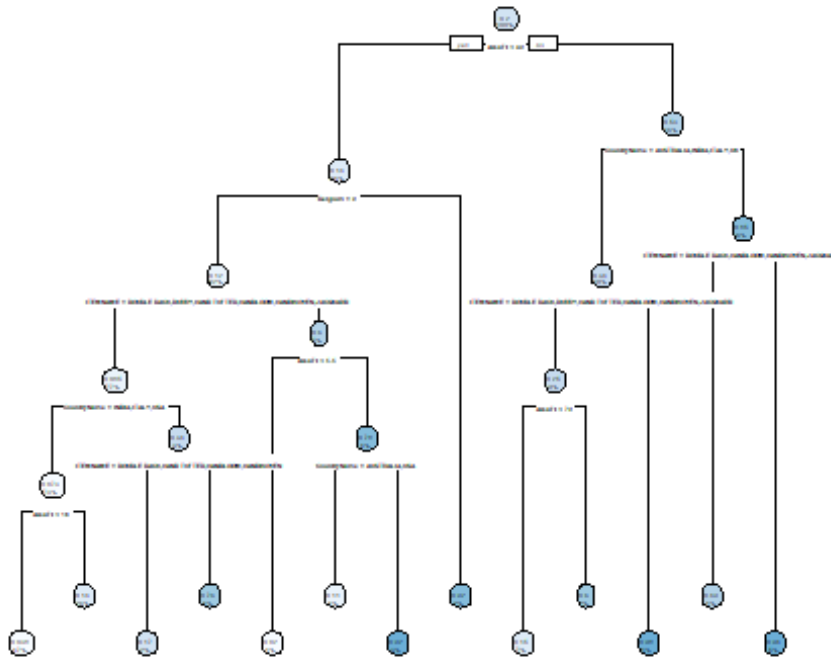
## [1] 0.7048953

myFormula = OrderConversion ~ .
myTree <- rpart(myFormula, data = train)
print(myTree)

## n= 4075
##
## node), split, n, deviance, yval
##      * denotes terminal node
##
## 1) root 4075 655.5912000 0.20147240
##    2) AreaFt< 39.90625 3423 408.3628000 0.13847500
##      4) Belgium< 0.5 3333 345.1311000 0.11731170
##        8) ITEMNAME=DOUBLE BACK,DURRY,HAND TUFTED,HANDLOOM,HANDWOVEN,JACQUA
RD 3121 241.6687000 0.08458827
##          16) CountryName=INDIA,ITALY,USA 2990 202.9595000 0.07324415
##            32) AreaFt< 15.875 1929 71.1612200 0.03836185 *
##            33) AreaFt>=15.875 1061 125.1838000 0.13666350 *
##          17) CountryName=AUSTRALIA,ROMANIA,UK 131 29.5419800 0.34351150
##            34) ITEMNAME=DOUBLE BACK,HAND TUFTED,HANDLOOM,HANDWOVEN 93 13.2
473100 0.17204300 *
##            35) ITEMNAME=DURRY,JACQUARD 38 6.8684210 0.76315790 *
##          9) ITEMNAME=GUN TUFTED,KNOTTED,POWER LOOM JACQUARD,TABLE TUFTED 212
50.9198100 0.59905660
##        18) AreaFt< 5.5 51 0.9803922 0.01960784 *
```

DESCISION TREE MODEL

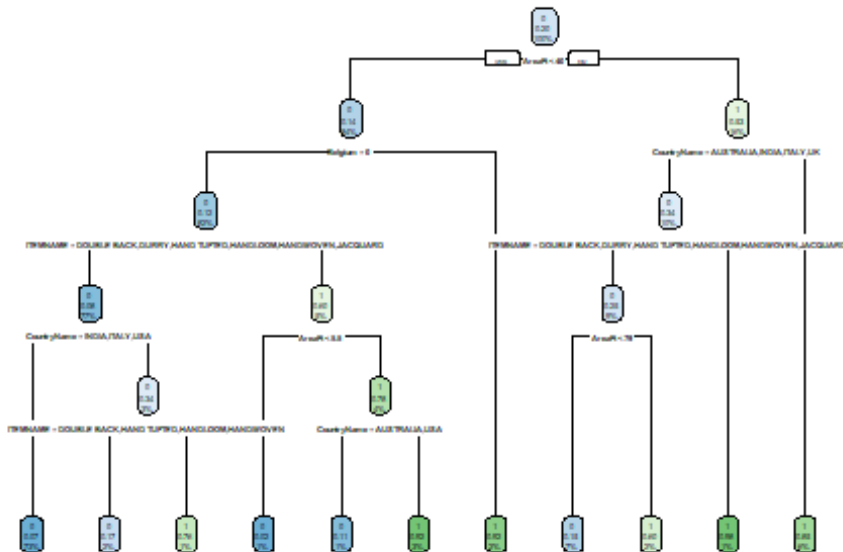
```
##          19) AreaFt>=5.5 161  27.3913000 0.78260870
##          38) CountryName=AUSTRALIA,USA 28  2.6785710 0.10714290 *
##          39) CountryName=INDIA,UK 133  9.2481200 0.92481200 *
##          5) Belgium>=0.5 90  6.4555560 0.92222220 *
##          3) AreaFt>=39.90625 652 162.3236000 0.53220860
##          6) CountryName=AUSTRALIA,INDIA,ITALY,UK 406  91.4113300 0.34236450
##          12) ITEMNAME=DOUBLE BACK,DURRY,HAND TUFTED,HANDLOOM,HANDWOVEN,JACQUA
RD 355  66.6873200 0.25070420
##          24) AreaFt< 79 275  34.8872700 0.14909090 *
##          25) AreaFt>=79 80  19.2000000 0.60000000 *
##          13) ITEMNAME=GUN TUFTED,KNOTTED,POWER LOOM JACQUARD 51  0.9803922 0
.98039220 *
##          7) CountryName=BELGIUM,ROMANIA,USA 246  32.1300800 0.84552850
##          14) ITEMNAME=DOUBLE BACK,HANDLOOM,HANDWOVEN,JACQUARD 51  12.7451000
0.49019610 *
##          15) ITEMNAME=DURRY,HAND TUFTED,KNOTTED 195  11.2615400 0.93846150 *
rpart.plot(myTree)
```



DECISION TREE MODEL

USING GINI INDEX

```
mytree <- rpart(myFormula, data = train, method="class", , parms=list(split="gini"), control = rpart.control(minsplit = 10, minbucket = 10), cp=0.03)
rpart.plot(mytree)
```



```
pred_train <- predict(mytree, data=train, type = "class", parms=list(split="gini"), control = rpart.control(minsplit = 10, minbucket = 10), cp=0.03)
mean(train$OrderConversion != pred_train)
```

```
## [1] 0.09251534
```

```
pred_test <- predict(mytree, newdata=test, type = "class", parms=list(split="gini"), control = rpart.control(minsplit = 10, minbucket = 10), cp=0.03)
mean(test$OrderConversion != pred_test)
```

```
## Warning in `!=.default`(test$OrderConversion, pred_test): longer object length is not a multiple of shorter object length
```

```
## is not a multiple of shorter object length
```

```
## Warning in is.na(e1) | is.na(e2): longer object length is not a multiple of shorter object length
```

```
## shorter object length
```

```
## [1] 0.2807362
```

DESCISION TREE MODEL

```
table_mat <- table(train$OrderConversion, pred_train)
table_mat

##      pred_train
##      0      1
## 0 3157    97
## 1  280   541

accuracy_Test <- sum(diag(table_mat)) / sum(table_mat)
print(paste('Accuracy for test', accuracy_Test))

## [1] "Accuracy for test 0.907484662576687"

printcp(mytree)

##
## Classification tree:
## rpart(formula = myFormula, data = train, method = "class", parms = list(sp
lit = "g\nini"),
##      control = rpart.control(minsplit = 10, minbucket = 10), cp = 0.03)
##
## Variables actually used in tree construction:
## [1] AreaFt      Belgium      CountryName ITEMNAME
##
## Root node error: 821/4075 = 0.20147
##
## n= 4075
##
##      CP nsplit rel error  xerror    xstd
## 1 0.103532      0  1.00000 1.00000 0.031187
## 2 0.092570      2  0.79294 0.79172 0.028470
## 3 0.059683      3  0.70037 0.70767 0.027186
## 4 0.055420      4  0.64068 0.65895 0.026383
## 5 0.026797      6  0.52984 0.53471 0.024107
## 6 0.019488      7  0.50305 0.50914 0.023591
## 7 0.012180      8  0.48356 0.50426 0.023491
## 8 0.010000     10  0.45920 0.50305 0.023465

printcp(mytree)

##
## Classification tree:
## rpart(formula = myFormula, data = train, method = "class", parms = list(sp
lit = "g\nini"),
##      control = rpart.control(minsplit = 10, minbucket = 10), cp = 0.03)
##
## Variables actually used in tree construction:
## [1] AreaFt      Belgium      CountryName ITEMNAME
##
## Root node error: 821/4075 = 0.20147
##
```

DESCISION TREE MODEL

```
## n= 4075
##
##      CP nsplit rel error  xerror    xstd
## 1 0.103532      0  1.00000 1.00000 0.031187
## 2 0.092570      2  0.79294 0.79172 0.028470
## 3 0.059683      3  0.70037 0.70767 0.027186
## 4 0.055420      4  0.64068 0.65895 0.026383
## 5 0.026797      6  0.52984 0.53471 0.024107
## 6 0.019488      7  0.50305 0.50914 0.023591
## 7 0.012180      8  0.48356 0.50426 0.023491
## 8 0.010000     10  0.45920 0.50305 0.023465
```

LOGISTIC REGRESSION

```
mydata = subset(mydata, select = -c(2))
head(mydata)

## # A tibble: 6 x 24
##   CustomerCode  USA    UK Italy Belgium Romania Australia India QtyRequir
ed
##   <chr>        <dbl> <dbl> <dbl>  <dbl>  <dbl>    <dbl> <dbl>    <dbl>
1>
## 1 A-11          1     0     0      0      0      0     0
2
## 2 A-11          1     0     0      0      0      0     0
2
## 3 A-11          1     0     0      0      0      0     0
2
## 4 A-11          1     0     0      0      0      0     0
1
## 5 A-9           1     0     0      0      0      0     0
5
## 6 A-9           1     0     0      0      0      0     0
5
## # ... with 15 more variables: ITEMNAME <chr>, HandTufted <dbl>, Durry <dbl>,
## #   DoubleBack <dbl>, HandWoven <dbl>, Knotted <dbl>, Jacquard <dbl>,
## #   Handloom <dbl>, Other <dbl>, ShapeName <chr>, REC <dbl>, Round <dbl>,
## #   Square <dbl>, AreaFt <dbl>, OrderConversion <dbl>

mydata$ITEMNAME <- as.numeric(factor(as.matrix(mydata$ITEMNAME)))
mydata$ShapeName <- as.numeric(factor(as.matrix(mydata$ShapeName)))
mydata$CustomerCode <- as.numeric(factor(as.matrix(mydata$CustomerCode)))

sapply(mydata, sd)

##   CustomerCode          USA          UK          Italy          Be
lgium
##   6.06258072      0.43252452      0.18408647      0.08789121      0.149
38512
```


DESCISION TREE MODEL

##	Romania	Australia	India	QtyRequired	ITE
MNAME					
##	0.05872169	0.04155853	0.46585067	5.70186587	2.102
92013					
##	HandTufted	Durry	DoubleBack	HandWoven	Kn
otted					
##	0.49320195	0.44310911	0.29343256	0.32725777	0.190
08945					
##	Jacquard	Handloom	Other	ShapeName	
REC					
##	0.11391193	0.13039081	0.16799129	0.15742051	0.116
10789					
##	Round	Square	AreaFt	OrderConversion	
##	0.09881476	0.06157713	21.58072448	0.39935834	

Converting the target variable into categorical variable

```
mydata$OrderConversion = factor(mydata$OrderConversion)
mylogit = glm(OrderConversion ~ ., data = mydata, family = "binomial")
summary(mylogit)
```

```
##
## Call:
## glm(formula = OrderConversion ~ ., family = "binomial", data = mydata)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -3.0003  -0.5838  -0.2655  -0.1937   2.9829
##
## Coefficients: (4 not defined because of singularities)
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -2.771625   1.504858  -1.842   0.0655 .
## CustomerCode -0.084170   0.009133  -9.216 < 2e-16 ***
## USA          1.950556   0.110686  17.622 < 2e-16 ***
## UK           3.114674   0.229502  13.571 < 2e-16 ***
## Italy        1.205744   0.528207   2.283   0.0224 *
## Belgium      7.172141   0.393417  18.230 < 2e-16 ***
## Romania      3.526021   0.507606   6.946 3.75e-12 ***
## Australia    -0.398053   0.834774  -0.477   0.6335
## India         NA         NA         NA      NA
## QtyRequired   0.012970   0.006547   1.981   0.0476 *
## ITEMNAME      0.245822   0.059973   4.099 4.15e-05 ***
## HandTufted    -3.137228   0.323028  -9.712 < 2e-16 ***
## Durry         -2.466272   0.421541  -5.851 4.90e-09 ***
## DoubleBack    -2.481866   0.496060  -5.003 5.64e-07 ***
## HandWoven     -4.341886   0.294290 -14.754 < 2e-16 ***
## Knotted       -1.440230   0.266355  -5.407 6.40e-08 ***
## Jacquard      -4.266561   0.404220 -10.555 < 2e-16 ***
## Handloom      -3.470214   0.401869  -8.635 < 2e-16 ***
## Other         NA         NA         NA      NA
```

DESCISION TREE MODEL

```
## ShapeName      1.116868    0.652405    1.712    0.0869 .
## REC            0.303105    0.890790    0.340    0.7337
## Round          NA          NA          NA          NA
## Square         NA          NA          NA          NA
## AreaFt         0.058131    0.002399    24.229    < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 5771.2  on 5780  degrees of freedom
## Residual deviance: 3637.7  on 5761  degrees of freedom
## AIC: 3677.7
##
## Number of Fisher Scoring iterations: 6
```

Confidence using standard errors

```
confint(mylogit)

## Waiting for profiling to be done...

##              2.5 %      97.5 %
## (Intercept) -5.686117076  0.25024041
## CustomerCode -0.102237222 -0.06642042
## USA          1.734398031  2.16843937
## UK           2.664437197  3.56480443
## Italy         0.122120250  2.21010343
## Belgium      6.440979573  7.99411018
## Romania      2.501229177  4.51413502
## Australia   -2.254510950  1.13814790
## India        NA          NA
## QtyRequired -0.000407948  0.02552246
## ITEMNAME     0.129111972  0.36510007
## HandTufted   -3.772908583 -2.49999503
## Durry        -3.295205445 -1.63350703
## DoubleBack   -3.457581395 -1.50382220
## HandWoven    -4.925722829 -3.76963709
## Knotted      -1.973329270 -0.92676053
## Jacquard     -5.095712313 -3.50430456
## Handloom     -4.272190943 -2.69308427
## Other        NA          NA
## ShapeName    -0.207331667  2.36942268
## REC          -1.447360027  2.06103259
## Round        NA          NA
## Square       NA          NA
## AreaFt       0.053499641  0.06290785

confint.default(mylogit)
```

DESCISION TREE MODEL

```
##              2.5 %      97.5 %
## (Intercept) -5.7210927484 0.17784195
## CustomerCode -0.1020704785 -0.06626957
## USA          1.7336156454 2.16749643
## UK           2.6648588133 3.56448967
## Italy         0.1704777114 2.24101016
## Belgium      6.4010575211 7.94322513
## Romania      2.5311320999 4.52091068
## Australia    -2.0341787328 1.23807358
## India        NA          NA
## QtyRequired  0.0001392121 0.02580140
## ITEMNAME     0.1282774504 0.36336659
## HandTufted   -3.7703506333 -2.50410506
## Durry        -3.2924779752 -1.64006656
## DoubleBack    -3.4541252522 -1.50960773
## HandWoven     -4.9186835743 -3.76508842
## Knotted       -1.9622750283 -0.91818438
## Jacquard      -5.0588162594 -3.47430484
## Handloom      -4.2578622515 -2.68256608
## Other         NA          NA
## ShapeName    -0.1618212298 2.39555746
## REC          -1.4428119266 2.04902117
## Round         NA          NA
## Square        NA          NA
## AreaFt        0.0534288937 0.06283371
```

Odd Ratio

```
exp(coef(mylogit))
```

```
## (Intercept) CustomerCode      USA      UK      Italy      Belg
ium
## 6.256024e-02 9.192749e-01 7.032597e+00 2.252609e+01 3.339242e+00 1.302631e
+03
##      Romania      Australia      India QtyRequired      ITEMNAME      HandTuf
ted
## 3.398847e+01 6.716267e-01      NA 1.013055e+00 1.278672e+00 4.340295e
-02
##      Durry      DoubleBack      HandWoven      Knotted      Jacquard      Handl
oom
## 8.490076e-02 8.358707e-02 1.301196e-02 2.368733e-01 1.402996e-02 3.111037e
-02
##      Other      ShapeName      REC      Round      Square      Are
aFt
##      NA 3.055270e+00 1.354056e+00      NA      NA 1.059854e
+00
```

```
exp(cbind(OR = coef(mylogit), confint(mylogit)))
```

```
## Waiting for profiling to be done...
```

DESCISION TREE MODEL

```
##              OR          2.5 %          97.5 %
## (Intercept)  6.256024e-02 3.392741e-03 1.284334e+00
## CustomerCode 9.192749e-01 9.028154e-01 9.357374e-01
## USA          7.032597e+00 5.665516e+00 8.744626e+00
## UK           2.252609e+01 1.435987e+01 3.533254e+01
## Italy         3.339242e+00 1.129890e+00 9.116659e+00
## Belgium      1.302631e+03 6.270207e+02 2.963452e+03
## Romania      3.398847e+01 1.219748e+01 9.129856e+01
## Australia    6.716267e-01 1.049248e-01 3.120983e+00
## India        NA          NA          NA
## QtyRequired  1.013055e+00 9.995921e-01 1.025851e+00
## ITEMNAME      1.278672e+00 1.137818e+00 1.440658e+00
## HandTufted    4.340295e-02 2.298511e-02 8.208541e-02
## Durry         8.490076e-02 3.706043e-02 1.952436e-01
## DoubleBack    8.358707e-02 3.150587e-02 2.222789e-01
## HandWoven     1.301196e-02 7.257478e-03 2.306043e-02
## Knotted       2.368733e-01 1.389933e-01 3.958339e-01
## Jacquard      1.402996e-02 6.122944e-03 3.006768e-02
## Handloom      3.111037e-02 1.395118e-02 6.767190e-02
## Other        NA          NA          NA
## ShapeName     3.055270e+00 8.127500e-01 1.069122e+01
## REC          1.354056e+00 2.351904e-01 7.854076e+00
## Round        NA          NA          NA
## Square        NA          NA          NA
## AreaFt        1.059854e+00 1.054957e+00 1.064929e+00
```

```
newdata1 = with(mydata, data.frame(CustomerCode = mean(CustomerCode),
                                   QtyRequired  = mean(QtyRequired),
                                   ITEM_NAME     = factor(1:12),
                                   ShapeName     = mean(ShapeName),
                                   AreaFt       = mean(AreaFt)))
```

newdata1

```
## CustomerCode QtyRequired ITEM_NAME ShapeName AreaFt
## 1      8.854869   1.977859         1  1.017471 21.63013
## 2      8.854869   1.977859         2  1.017471 21.63013
## 3      8.854869   1.977859         3  1.017471 21.63013
## 4      8.854869   1.977859         4  1.017471 21.63013
## 5      8.854869   1.977859         5  1.017471 21.63013
## 6      8.854869   1.977859         6  1.017471 21.63013
## 7      8.854869   1.977859         7  1.017471 21.63013
## 8      8.854869   1.977859         8  1.017471 21.63013
## 9      8.854869   1.977859         9  1.017471 21.63013
## 10     8.854869   1.977859        10  1.017471 21.63013
## 11     8.854869   1.977859        11  1.017471 21.63013
## 12     8.854869   1.977859        12  1.017471 21.63013
```

```
Pred <- predict(mylogit, newdata2 = newdata1, type = "response")
```

DESCISION TREE MODEL

Difference between deviance for the two models = test statistic

```
with(mylogit, null.deviance - deviance)
## [1] 2133.524
```

The degrees of freedom for the difference between the two models is equal to the number of predictor variables in the model, and can be obtained using:

```
with(mylogit, df.null, df.residual)
## [1] 5780
```

Finally, the p-value can be obtained using

```
with(mylogit, pchisq(null.deviance - deviance, df.null - df.residual, lower.tail = FALSE))
## [1] 0
class <- ifelse(Pred >= 0.5, "YES", "NO")
```

Accuracy using the Decision Tree Model is around 90.74%

--END OF DECISION TREE MODEL--

CHAMPO CARPETS – RANDOM FOREST MODEL

```

Sampledata <- read_excel("C:/Users/Bodake/Desktop/UIC Sem 2/Data Mining IDS-5
72/R/sample.xlsx")
View(Sampledata)
mydata <- Sampledata

str(mydata)

## tibble [5,820 x 25] (S3: tbl_df/tbl/data.frame)
## $ CustomerCode : chr [1:5820] "A-11" "A-11" "A-11" "A-11" ...
## $ CountryName : chr [1:5820] "USA" "USA" "USA" "USA" ...
## $ USA : num [1:5820] 1 1 1 1 1 1 1 1 1 1 ...
## $ UK : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Italy : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Belgium : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Romania : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Australia : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ India : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ QtyRequired : num [1:5820] 2 2 2 1 5 5 5 5 11 2 ...
## $ ITEMNAME : chr [1:5820] "DURRY" "DOUBLE BACK" "JACQUARD" "JACQUAR
D" ...
## $ HandTufted : num [1:5820] 0 0 0 0 0 0 0 0 0 1 ...
## $ Durry : num [1:5820] 1 0 0 0 0 0 0 0 0 0 ...
## $ DoubleBack : num [1:5820] 0 1 0 0 1 1 1 1 0 0 ...
## $ HandWoven : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Knotted : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Jacquard : num [1:5820] 0 0 1 1 0 0 0 0 1 0 ...
## $ Handloom : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Other : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ ShapeName : chr [1:5820] "REC" "REC" "REC" "REC" ...
## $ REC : num [1:5820] 1 1 1 1 1 1 1 1 1 1 ...
## $ Round : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ Square : num [1:5820] 0 0 0 0 0 0 0 0 0 0 ...
## $ AreaFt : num [1:5820] 4 4 4 42.3 1 ...
## $ OrderConversion: num [1:5820] 0 0 0 1 0 0 0 0 0 0 ...

sum(is.na(mydata))

## [1] 273

mydata[is.na(mydata)] = 0
sum(is.na(mydata))

## [1] 0

sapply(mydata,function(x)sum(is.na(x)))

```

RANDOM FOREST MODEL

```
##      CustomerCode      CountryName      USA      UK
Italy
##          0          0          0          0
0
##      Belgium      Romania      Australia      India      QtyReq
uired
##          0          0          0          0
0
##      ITEMNAME      HandTufted      Durry      DoubleBack      Hand
Woven
##          0          0          0          0
0
##      Knotted      Jacquard      Handloom      Other      Shap
eName
##          0          0          0          0
0
##      REC      Round      Square      AreaFt      OrderConve
rsion
##          0          0          0          0
0
```

```
mydata <- as.data.frame(mydata)
mydata <- subset (mydata, select = -c(USA, UK, Italy, Belgium, Romania, Austr
alia, India))
str(mydata)

mydata$OrderConversion <-as.factor(factor(as.matrix(mydata$OrderConversion)))
```

#Random Forest Model

```
set.seed(100)
rf <- randomForest(mydata$OrderConversion ~ ., data = mydata, ntree = 300, mtr
y
                    = sqrt(ncol(mydata)-1), proximity = T, importance = T )
print(rf)

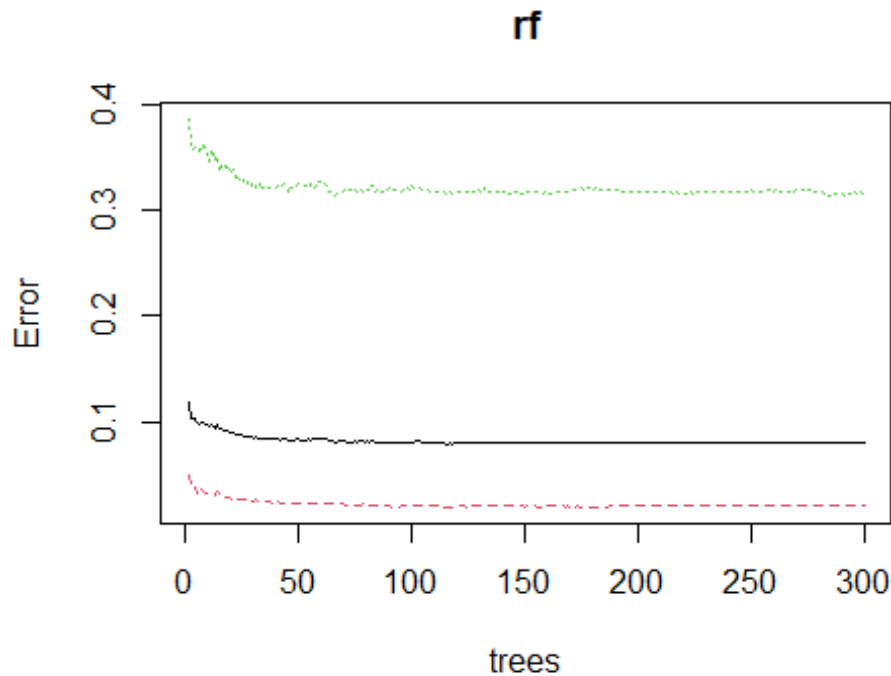
##
## Call:
## randomForest(formula = mydata$OrderConversion ~ ., data = mydata, ntree = 300, mtry = sqrt(ncol(mydata) - 1), proximity = T, importance = T)
##      Type of random forest: classification
##      Number of trees: 300
## No. of variables tried at each split: 4
##
##      OOB estimate of error rate: 8.09%
## Confusion matrix:
##      0  1 class.error
## 0 4550 101  0.02171576
## 1  370 799  0.31650984
```

RANDOM FOREST MODEL

```
attributes(rf)

## $names
## [1] "call"          "type"          "predicted"     "err.rate"
## [5] "confusion"     "votes"         "oob.times"     "classes"
## [9] "importance"    "importanceSD"  "localImportance" "proximity"
## [13] "ntree"         "mtry"          "forest"        "y"
## [17] "test"         "inbag"         "terms"
##
## $class
## [1] "randomForest.formula" "randomForest"

plot(rf)
```



```
p1 <- predict(rf, mydata)
confusionMatrix(p1, mydata$OrderConversion)

## Confusion Matrix and Statistics
##
##           Reference
## Prediction    0    1
##           0 4585  325
##           1   66  844
##
##           Accuracy : 0.9328
##           95% CI : (0.9261, 0.9391)
##           No Information Rate : 0.7991
```


RANDOM FOREST MODEL

```
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.7718
##
## Mcnemar's Test P-Value : < 2.2e-16
##
##      Sensitivity : 0.9858
##      Specificity : 0.7220
##      Pos Pred Value : 0.9338
##      Neg Pred Value : 0.9275
##      Prevalence : 0.7991
##      Detection Rate : 0.7878
##      Detection Prevalence : 0.8436
##      Balanced Accuracy : 0.8539
##
##      'Positive' Class : 0
##
```

Hence, we can say that the ACCURACY OF RANDOM FOREST MODEL ON OUR DATA SET IS Around 93.3%

```
rf$importance
##              0              1 MeanDecreaseAccuracy MeanDecreaseGi
ni
## CustomerCode 0.0646396504 0.0954731843          0.0708230649      137.6328
50
## CountryName  0.0604618670 0.1396365557          0.0764096629      227.2380
58
## QtyRequired  0.0126190009 0.0187168546          0.0138396952       59.2091
02
## ITEMNAME     0.0317481579 0.1123979410          0.0480179342      158.8418
82
## HandTufted   0.0099520814 0.0392259754          0.0158561771       20.8640
11
## Durry        0.0073486691 0.0163196586          0.0091820266       12.9136
27
## DoubleBack   0.0028084925 0.0031394463          0.0028756082        6.7083
94
## HandWoven    0.0036159754 0.0130353377          0.0055149885       10.8905
96
## Knotted      0.0063744450 0.0297972179          0.0110864751       36.4349
19
## Jacquard     0.0007098712 0.0001888175          0.0006036450        3.0705
36
## Handloom     0.0011495385 0.0016574721          0.0012533988        8.3225
12
## Other        0.0060933853 0.0284706373          0.0105960845       53.9485
06
## ShapeName    0.0006676833 0.0016274978          0.0008615113        4.3851
```

RANDOM FOREST MODEL

```
76
## REC          0.0008059448 0.0017874371          0.0010039013          3.7482
23
## Round        0.0002526424 0.0010040740          0.0004028903          2.2161
22
## Square       0.0002057407 0.0006069963          0.0002872994          2.2244
54
## AreaFt       0.0537991030 0.2040063657          0.0841027039          350.2552
06
```

```
importance(rf, type = 1)
```

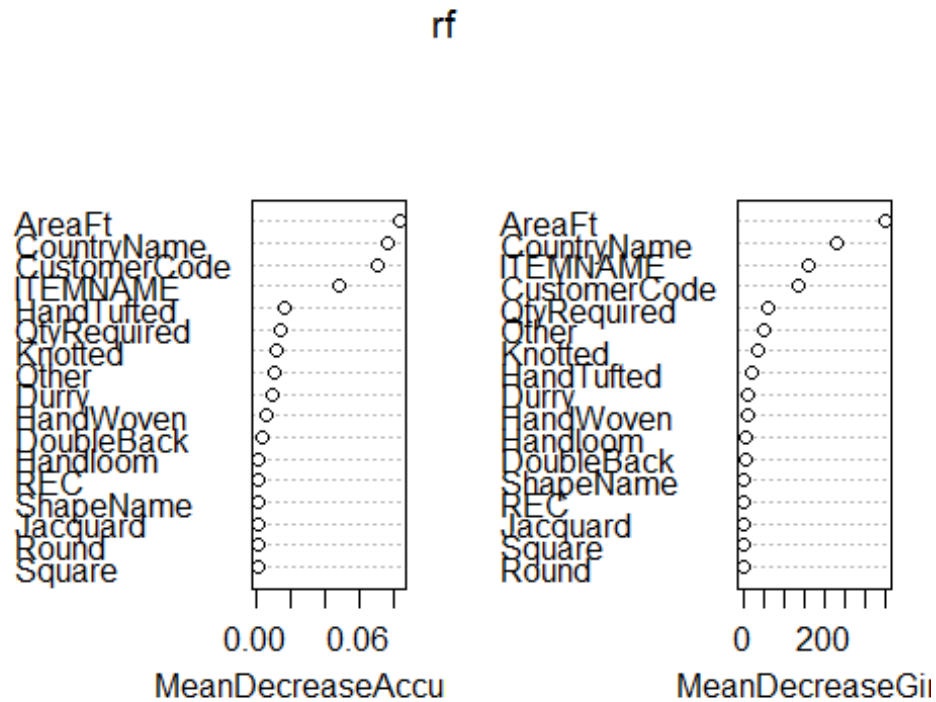
```
##              MeanDecreaseAccuracy
## CustomerCode      25.388151
## CountryName       41.369526
## QtyRequired       35.330187
## ITEMNAME          32.495191
## HandTufted        19.806019
## Durry             12.171852
## DoubleBack        10.019951
## HandWoven         13.239589
## Knotted           13.745132
## Jacquard           7.663893
## Handloom          14.064707
## Other             16.796032
## ShapeName         10.236109
## REC               9.808721
## Round             6.955230
## Square            7.489018
## AreaFt            71.507265
```

```
importance(rf, type = 2)
```

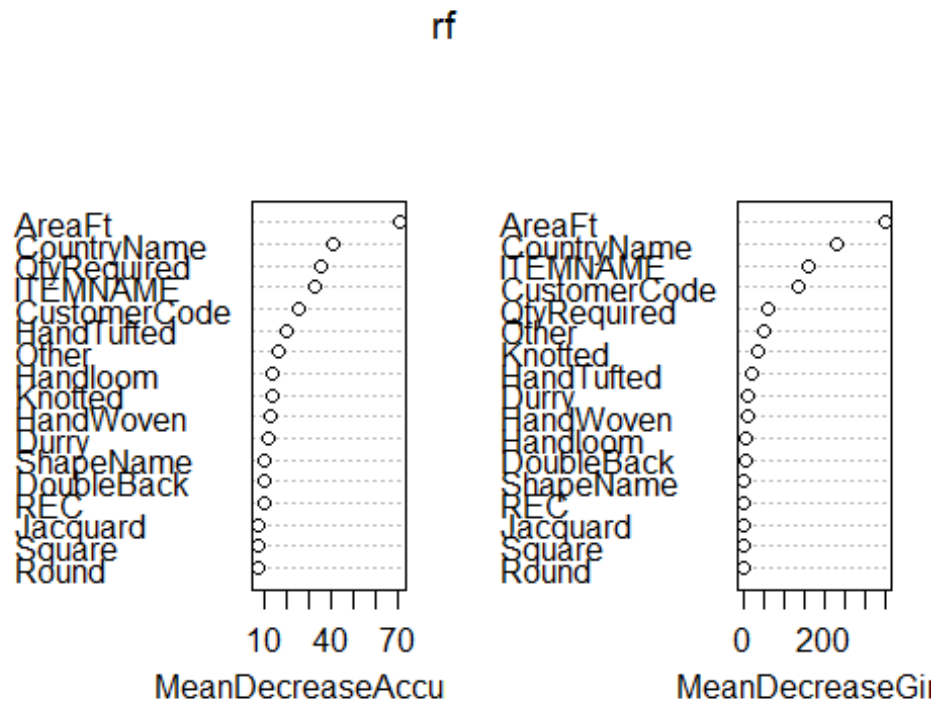
```
##              MeanDecreaseGini
## CustomerCode      137.632850
## CountryName       227.238058
## QtyRequired       59.209102
## ITEMNAME          158.841882
## HandTufted        20.864011
## Durry             12.913627
## DoubleBack        6.708394
## HandWoven         10.890596
## Knotted           36.434919
## Jacquard           3.070536
## Handloom          8.322512
## Other            53.948506
## ShapeName         4.385176
## REC               3.748223
## Round            2.216122
## Square            2.224454
## AreaFt           350.255206
```

RANDOM FOREST MODEL

```
varImpPlot(rf, scale=FALSE)
```



```
varImpPlot(rf)
```



RANDOM FOREST MODEL

```
## [299,] 0.08041237 0.02128575 0.3156544
## [300,] 0.08092784 0.02171576 0.3165098

ntree=100
rf$err.rate[ntree,1]

##          OOB
## 0.08127148

head(rf$predicted)

## 1 2 3 4 5 6
## 0 0 0 0 0 0
## Levels: 0 1

ind <- sample(2, nrow(mydata), replace = T, prob = c(0.7, 0.3))
Train <- mydata[ind == 1, ]
Validation <- mydata[ind == 2, ]
pr.err <- c()

rf$predicted

## Levels: 0 1

table(rf$predicted, mydata$OrderConversion, dnn = c("Predicted", "Actual"))

##          Actual
## Predicted    0    1
##           0 4550  370
##           1  101  799

confusionMatrix(rf$predicted, mydata$OrderConversion)

## Confusion Matrix and Statistics
##
##              Reference
## Prediction    0    1
##           0 4550  370
##           1  101  799
##
##              Accuracy : 0.9191
##              95% CI : (0.9118, 0.926)
##      No Information Rate : 0.7991
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.7242
##
##  Mcnemar's Test P-Value : < 2.2e-16
##
##              Sensitivity : 0.9783
##              Specificity : 0.6835
```

RANDOM FOREST MODEL

```
##          Pos Pred Value : 0.9248
##          Neg Pred Value : 0.8878
##          Prevalence     : 0.7991
##          Detection Rate  : 0.7818
##          Detection Prevalence : 0.8454
##          Balanced Accuracy : 0.8309
##
##          'Positive' Class : 0
##

library(ROCR)

pred <- prediction(rf$votes[,2], mydata$OrderConversion)

pred

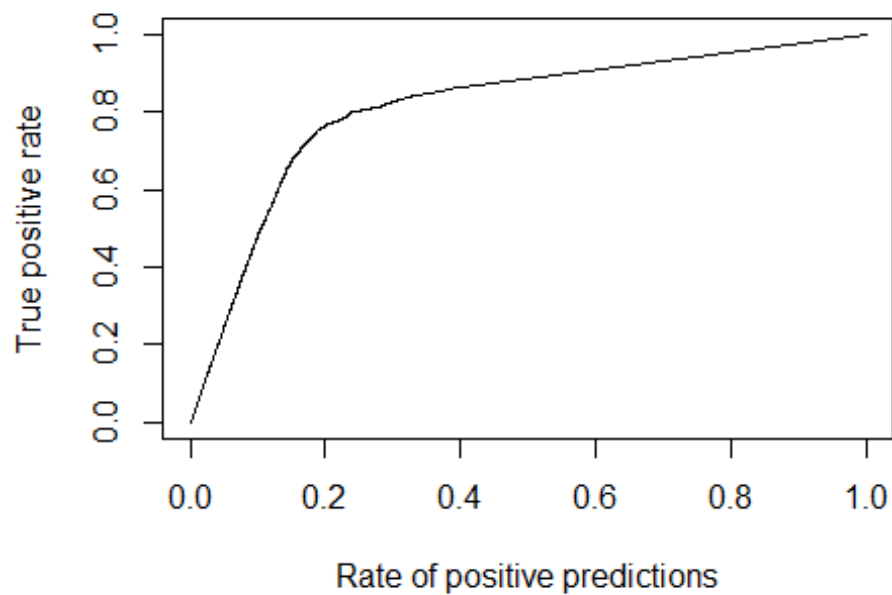
## A prediction instance
##   with 5820 data points

#Gain Chart
per <- performance(pred, "tpr", "rpp")
per

## A performance instance
##   'Rate of positive predictions' vs. 'True positive rate' (alpha: 'Cutoff'
## )
##   with 1191 data points

plot(per)
```

RANDOM FOREST MODEL

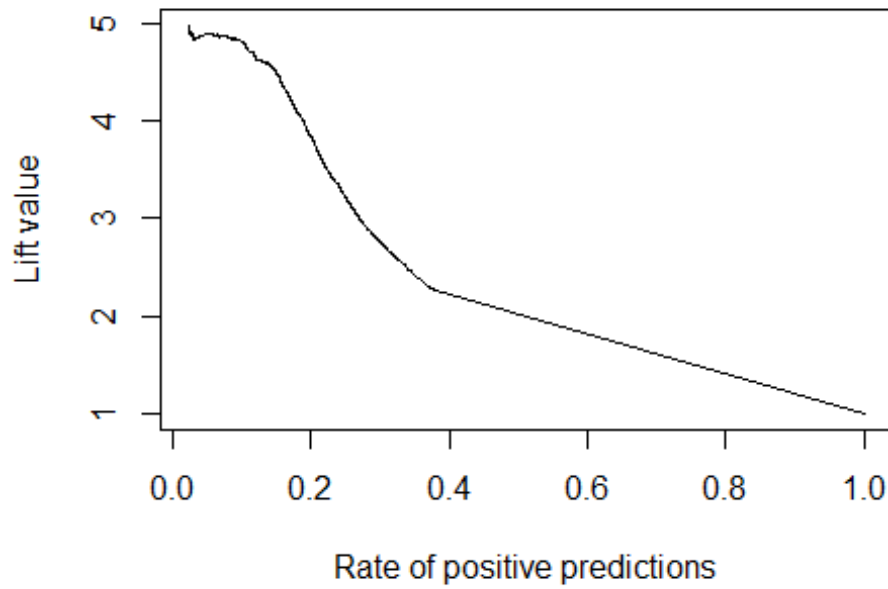


```
#Lift Chart
perf <- performance(pred, "lift", "rpp")
perf

## A performance instance
## 'Rate of positive predictions' vs. 'Lift value' (alpha: 'Cutoff')
## with 1191 data points

plot(perf)
```

RANDOM FOREST MODEL



#Response Chart

```
perfo <- performance(pred, "ppv", "rpp")  
perfo
```

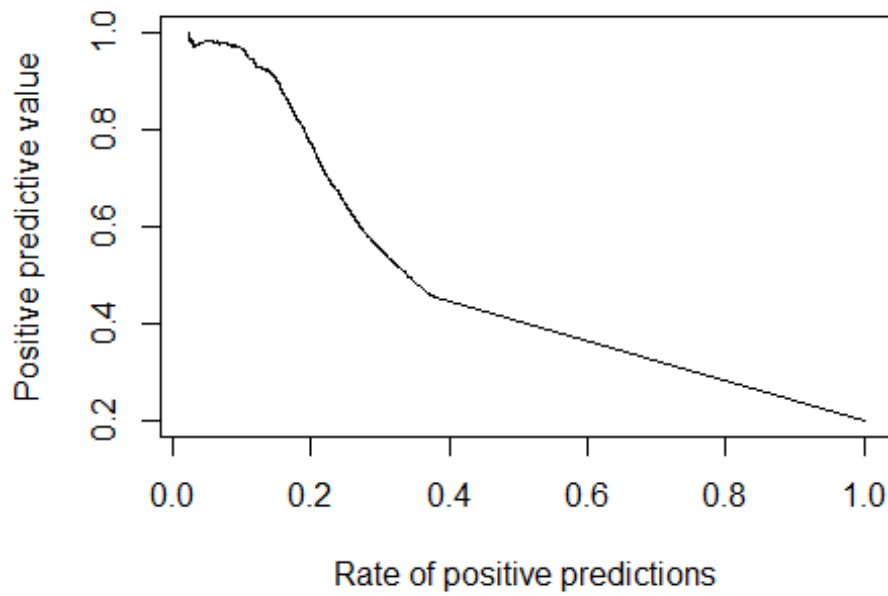
```
## A performance instance
```

```
## 'Rate of positive predictions' vs. 'Positive predictive value' (alpha: '  
Cutoff')
```

```
## with 1191 data points
```

```
plot(perfo)
```

RANDOM FOREST MODEL



#ROC Curve

```
perform <- performance(pred, "tpr", "fpr")  
perform
```

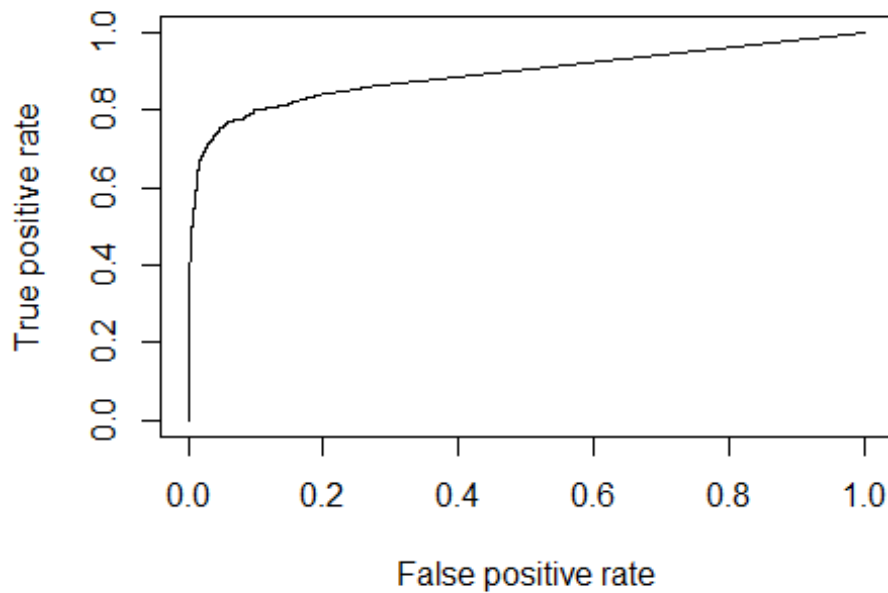
```
## A performance instance
```

```
## 'False positive rate' vs. 'True positive rate' (alpha: 'Cutoff')
```

```
## with 1191 data points
```

```
plot(perform)
```


RANDOM FOREST MODEL



```
#AUC
auc <- performance(pred, "auc")
auc

## A performance instance
## 'Area under the ROC curve'

unlist(slot(auc, "y.values"))

## [1] 0.8915729
```

Hence, we can say that the ACCURACY OF RANDOM FOREST MODEL ON OUR DATA SET IS Around 93.3%

---END OF RANDOM FOREST MODEL ANALYSIS---

CHAMPO CARPETS: LOGISTIC REGRESSION MODEL ON SAMPLE DATA

```

mydata <- read_excel("C:/Users/Bodake/Desktop/UIC Sem 2/Data Mining IDS-
572/R/sample.xlsx")
mydata <- na.omit(mydata)
mydata <- drop_na(mydata)

mydata = subset(mydata, select = -c(2))
head(mydata)

## # A tibble: 6 x 24
##   CustomerCode  USA    UK Italy Belgium Romania Australia India
##   <chr>          <dbl> <dbl> <dbl>   <dbl>   <dbl>   <dbl> <dbl>
##   <dbl>
## 1 A-11          1     0     0     0     0     0     0
## 2
## 2 A-11          1     0     0     0     0     0     0
## 2
## 3 A-11          1     0     0     0     0     0     0
## 2
## 4 A-11          1     0     0     0     0     0     0
## 1
## 5 A-9           1     0     0     0     0     0     0
## 5
## 6 A-9           1     0     0     0     0     0     0
## 5
## # ... with 15 more variables: ITEMNAME <chr>, HandTufted <dbl>, Durry
<dbl>,
## #   DoubleBack <dbl>, HandWoven <dbl>, Knotted <dbl>, Jacquard <dbl>,
## #   Handloom <dbl>, Other <dbl>, ShapeName <chr>, REC <dbl>, Round <dbl>,
## #   Square <dbl>, AreaFt <dbl>, OrderConversion <dbl>

mydata$ITEMNAME <- as.numeric(factor(as.matrix(mydata$ITEMNAME)))
mydata$ShapeName <- as.numeric(factor(as.matrix(mydata$ShapeName)))
mydata$CustomerCode <- as.numeric(factor(as.matrix(mydata$CustomerCode)))
str(mydata)

## tibble [5,781 x 24] (S3: tbl_df/tbl/data.frame)
## $ CustomerCode : num [1:5781] 1 1 1 1 2 2 2 2 2 2 ...
## $ USA          : num [1:5781] 1 1 1 1 1 1 1 1 1 1 ...
## $ UK           : num [1:5781] 0 0 0 0 0 0 0 0 0 0 ...
## $ Italy        : num [1:5781] 0 0 0 0 0 0 0 0 0 0 ...
## $ Belgium      : num [1:5781] 0 0 0 0 0 0 0 0 0 0 ...
## $ Romania      : num [1:5781] 0 0 0 0 0 0 0 0 0 0 ...
## $ Australia    : num [1:5781] 0 0 0 0 0 0 0 0 0 0 ...
## $ India        : num [1:5781] 0 0 0 0 0 0 0 0 0 0 ...
## $ QtyRequired  : num [1:5781] 2 2 2 1 5 5 5 5 11 2 ...

```

LOGISTIC REGRESSION MODEL

```
## $ ITEMNAME      : num [1:5781] 2 1 8 8 1 1 1 1 8 4 ...
## $ HandTufted    : num [1:5781] 0 0 0 0 0 0 0 0 0 1 ...
## $ Durry         : num [1:5781] 1 0 0 0 0 0 0 0 0 0 ...
## $ DoubleBack     : num [1:5781] 0 1 0 0 1 1 1 1 0 0 ...
## $ HandWoven      : num [1:5781] 0 0 0 0 0 0 0 0 0 0 ...
## $ Knotted        : num [1:5781] 0 0 0 0 0 0 0 0 0 0 ...
## $ Jacquard        : num [1:5781] 0 0 1 1 0 0 0 0 1 0 ...
## $ Handloom       : num [1:5781] 0 0 0 0 0 0 0 0 0 0 ...
## $ Other          : num [1:5781] 0 0 0 0 0 0 0 0 0 0 ...
## $ ShapeName      : num [1:5781] 1 1 1 1 1 1 1 1 1 1 ...
## $ REC            : num [1:5781] 1 1 1 1 1 1 1 1 1 1 ...
## $ Round          : num [1:5781] 0 0 0 0 0 0 0 0 0 0 ...
## $ Square         : num [1:5781] 0 0 0 0 0 0 0 0 0 0 ...
## $ AreaFt         : num [1:5781] 4 4 4 42.3 1 ...
## $ OrderConversion: num [1:5781] 0 0 0 1 0 0 0 0 0 0 ...
## - attr(*, "na.action")= 'omit' Named int [1:39] 227 4284 4285 4286 4287
4288 4289 4290 4291 4292 ...
## ... attr(*, "names")= chr [1:39] "227" "4284" "4285" "4286" ...
```

```
sapply(mydata, sd)
```

CustomerCode	USA	UK	Italy
Belgium			
6.06258072	0.43252452	0.18408647	0.08789121
0.14938512			
Romania	Australia	India	QtyRequired
ITEMNAME			
0.05872169	0.04155853	0.46585067	5.70186587
2.10292013			
HandTufted	Durry	DoubleBack	HandWoven
Knotted			
0.49320195	0.44310911	0.29343256	0.32725777
0.19008945			
Jacquard	Handloom	Other	ShapeName
REC			
0.11391193	0.13039081	0.16799129	0.15742051
0.11610789			
Round	Square	AreaFt	OrderConversion
0.09881476	0.06157713	21.58072448	0.39935834

Converting the target variable into categorical variable

```
mydata$OrderConversion = factor(mydata$OrderConversion)
mylogit = glm(OrderConversion ~ ., data = mydata, family = "binomial")
summary(mylogit)

##
## Call:
## glm(formula = OrderConversion ~ ., family = "binomial", data = mydata)
##
## Deviance Residuals:
```

LOGISTIC REGRESSION MODEL

```
##      Min      1Q   Median      3Q      Max
## -3.0003  -0.5838  -0.2655  -0.1937   2.9829
##
## Coefficients: (4 not defined because of singularities)
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -2.771625   1.504858  -1.842   0.0655 .
## CustomerCode -0.084170   0.009133  -9.216 < 2e-16 ***
## USA          1.950556   0.110686  17.622 < 2e-16 ***
## UK           3.114674   0.229502  13.571 < 2e-16 ***
## Italy        1.205744   0.528207   2.283   0.0224 *
## Belgium      7.172141   0.393417  18.230 < 2e-16 ***
## Romania      3.526021   0.507606   6.946 3.75e-12 ***
## Australia    -0.398053   0.834774  -0.477   0.6335
## India        NA         NA         NA      NA
## QtyRequired   0.012970   0.006547   1.981   0.0476 *
## ITEMNAME      0.245822   0.059973   4.099 4.15e-05 ***
## HandTufted    -3.137228   0.323028  -9.712 < 2e-16 ***
## Durry         -2.466272   0.421541  -5.851 4.90e-09 ***
## DoubleBack    -2.481866   0.496060  -5.003 5.64e-07 ***
## HandWoven     -4.341886   0.294290 -14.754 < 2e-16 ***
## Knotted       -1.440230   0.266355  -5.407 6.40e-08 ***
## Jacquard      -4.266561   0.404220 -10.555 < 2e-16 ***
## Handloom      -3.470214   0.401869  -8.635 < 2e-16 ***
## Other         NA         NA         NA      NA
## ShapeName     1.116868   0.652405   1.712   0.0869 .
## REC           0.303105   0.890790   0.340   0.7337
## Round         NA         NA         NA      NA
## Square        NA         NA         NA      NA
## AreaFt        0.058131   0.002399  24.229 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 5771.2  on 5780  degrees of freedom
## Residual deviance: 3637.7  on 5761  degrees of freedom
## AIC: 3677.7
##
## Number of Fisher Scoring iterations: 6
```

Performing ANOVA to analyze table of deviance

```
anova(mylogit, test="Chisq")

## Analysis of Deviance Table
##
## Model: binomial, link: logit
##
## Response: OrderConversion
##
```

LOGISTIC REGRESSION MODEL

```
## Terms added sequentially (first to last)
##
##
##           Df Deviance Resid. Df Resid. Dev  Pr(>Chi)
## NULL                                5780      5771.2
## CustomerCode  1      54.52      5779      5716.7 1.537e-13 ***
## USA           1      32.82      5778      5683.9 1.010e-08 ***
## UK            1      19.97      5777      5663.9 7.859e-06 ***
## Italy         1       0.40      5776      5663.5 0.5269861
## Belgium      1     471.78      5775      5191.7 < 2.2e-16 ***
## Romania      1      14.92      5774      5176.8 0.0001124 ***
## Australia    1       0.91      5773      5175.9 0.3413791
## India        0       0.00      5773      5175.9
## QtyRequired  1       4.98      5772      5170.9 0.0256835 *
## ITEMNAME     1     327.88      5771      4843.0 < 2.2e-16 ***
## HandTufted   1       0.11      5770      4842.9 0.7374292
## Durry        1       6.91      5769      4836.0 0.0085843 **
## DoubleBack   1     159.91      5768      4676.1 < 2.2e-16 ***
## HandWoven    1     101.80      5767      4574.3 < 2.2e-16 ***
## Knotted      1       0.49      5766      4573.8 0.4824902
## Jacquard     1      55.01      5765      4518.8 1.199e-13 ***
## Handloom     1      45.19      5764      4473.6 1.792e-11 ***
## Other        0       0.00      5764      4473.6
## ShapeName    1       5.10      5763      4468.5 0.0239327 *
## REC          1       1.05      5762      4467.5 0.3054274
## Round        0       0.00      5762      4467.5
## Square       0       0.00      5762      4467.5
## AreaFt       1     829.77      5761      3637.7 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

fitted.results <- predict(mylogit,mydata)

fitted.results <- ifelse(fitted.results > 0.5,1,0)

misClasificError <- mean(fitted.results != mydata$OrderConversion)
print(paste('Accuracy',1-misClasificError))

## [1] "Accuracy 0.877183878221761"
```

Hence, we can conclude that the accuracy of Logistic Regression Model on CHAMPO Sample dataset is around 88%

--END OF LOGISTIC REGRESSION MODEL--

NEURAL NETWORK MODEL

CHAMPO CARPETS: NEURAL NETWORK MODEL ON SAMPLE DATA

```
Data <- read_excel("C:/Users/Bodake/Desktop/UIC Sem 2/Data Mining IDS-572/R/sample.xlsx")
sum(is.na(Data)) #checking for NA's

## [1] 273

Data <- na_mean(Data)#Replacing the NA's with mean
sum(is.na(Data)) #checking if NA's were substituted correctly

## [1] 0

Data$Target <- as.factor(ifelse(Data$OrderConversion>0, "Conversion", "No_Conversion"))
df <- as.data.frame(Data)
df <- subset (df, select = -c(USA, UK, Italy, Belgium, Romania, Australia, India,OrderConversion)) #Removing countries as they are already included in `CountryName` column.
str(df)

## 'data.frame':    5820 obs. of  18 variables:
##  $ CustomerCode: chr  "A-11" "A-11" "A-11" "A-11" ...
##  $ CountryName : chr  "USA" "USA" "USA" "USA" ...
##  $ QtyRequired : num  2 2 2 1 5 5 5 5 11 2 ...
##  $ ITEMNAME     : chr  "DURRY" "DOUBLE BACK" "JACQUARD" "JACQUARD" ...
##  $ HandTufted   : num  0 0 0 0 0 0 0 0 0 1 ...
##  $ Durry        : num  1 0 0 0 0 0 0 0 0 0 ...
##  $ DoubleBack   : num  0 1 0 0 1 1 1 1 0 0 ...
##  $ HandWoven    : num  0 0 0 0 0 0 0 0 0 0 ...
##  $ Knotted      : num  0 0 0 0 0 0 0 0 0 0 ...
##  $ Jacquard      : num  0 0 1 1 0 0 0 0 1 0 ...
##  $ Handloom     : num  0 0 0 0 0 0 0 0 0 0 ...
##  $ Other        : num  0 0 0 0 0 0 0 0 0 0 ...
##  $ ShapeName    : chr  "REC" "REC" "REC" "REC" ...
##  $ REC          : num  1 1 1 1 1 1 1 1 1 1 ...
##  $ Round        : num  0 0 0 0 0 0 0 0 0 0 ...
##  $ Square       : num  0 0 0 0 0 0 0 0 0 0 ...
##  $ AreaFt       : num  4 4 4 42.3 1 ...
##  $ Target       : Factor w/ 2 levels "Conversion","No_Conversion": 2 2 2 1 2 2 2 2 2 2 ...
```

Now that we have cleaned our sample data by replacing the null values and removing redundant data, we can get started with implementing Neural Network model on our dataset

```
myscale <- function(x) {
  (x - min(x)) / (max(x) - min(x))
}
```

NEURAL NETWORK MODEL

```
}

df <- df %>% mutate_if(is.numeric, myscale)
#categorical variables remain untouched

set.seed(1234)
df$CustomerCode<- as.numeric(factor(as.matrix(df$CustomerCode)))
df$CountryName<- as.numeric(factor(as.matrix(df$CountryName)))
df$ITEM_NAME<- as.numeric(factor(as.matrix(df$ITEMNAME)))
df$ShapeName<- as.numeric(factor(as.matrix(df$ShapeName)))

ind <- sample(2, nrow(df), replace = T, prob = c(0.7, 0.3))
train <- df[ind == 1, ]
test <- df[ind == 2, ]

#Neural Net Model

#str(df)
#summary(df)

sapply(lapply(df, unique), length)

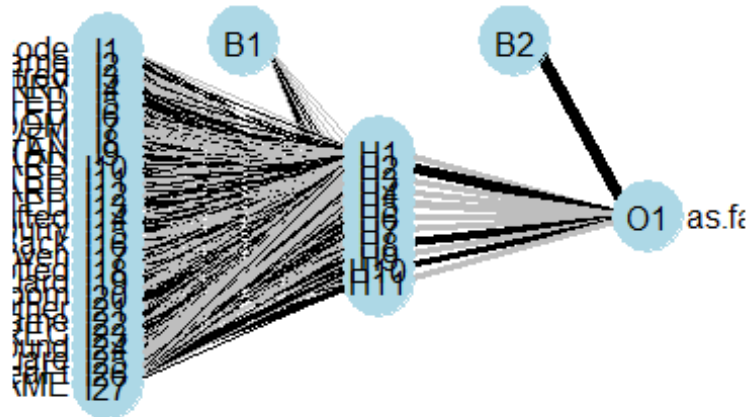
## CustomerCode CountryName QtyRequired ITEMNAME HandTufted Du
rry
## 34 14 41 11 2
2
## DoubleBack HandWoven Knotted Jacquard Handloom Ot
her
## 2 2 2 2 2
2
## ShapeName REC Round Square AreaFt Tar
get
## 3 2 2 2 139
2
## ITEM_NAME
## 11

neuralnModel <- nnet(as.factor(Target) ~ ., data = train,hidden=3 ,linout = F
,size = 11, decay = 0.01,
maxit = 1000)

## # weights: 320
## initial value 2187.883744
## iter 10 value 1780.107594
## iter 20 value 1672.149351
## iter 30 value 1578.914927
## iter 40 value 1474.603456
## iter 50 value 1302.278617
```

NEURAL NETWORK MODEL

```
## iter 60 value 1262.190067
## iter 70 value 1235.031910
## iter 80 value 1206.816505.....
.....
## stopped after 1000 iterations
plotnet(neuralnModel)
```



```
neuraln.preds = predict(neuralnModel, test)
neuraln.preds.class = as.factor(predict(neuralnModel, test, type = "class"))
#neuraln.preds.class
CM <- table(neuraln.preds.class, test$Target, dnn = c("predicted","actual"))
print(CM)
```

```
##           actual
## predicted  Conversion No_Conversion
## Conversion      239         35
## No_Conversion  106        1334
```

We print our Confusion matrix and observe the values

```
error_metric = function(CM)
{
  TN = CM[1,1]
  TP = CM[2,2]
  FN = CM[1,2]
  FP = CM[2,1]
  recall = (TP)/(TP+FN)
```


NEURAL NETWORK MODEL

```
precision =(TP)/(TP+FP)
falsePositiveRate = (FP)/(FP+TN)
falseNegativeRate = (FN)/(FN+TP)
error =(FP+FN)/(TP+TN+FP+FN)
modelPerf <- list("precision" = precision,
                  "recall" = recall,
                  "falsepositiverate" = falsePositiveRate,
                  "falsenegativerate" = falseNegativeRate,
                  "error" = error)

return(modelPerf)
}
outPutlist <- error_metric(CM)
library(plyr)

df <- ldply(outPutlist, data.frame)
setNames(df, c("", "Values"))

##              Values
## 1      precision 0.92638889
## 2          recall 0.97443389
## 3 falsepositiverate 0.30724638
## 4 falsenegativerate 0.02556611
## 5              error 0.08226371
```

Now that we have successfully executed the Neural Network model on our dataset we get the following information

Precision is about 92.31% Recall is around 97.37% FPR is around 32.1% FNR is around 2.62% and Error rate is around 8.57%

We can calculate Accuracy as 1-error i.e., $1 - 0.082 = 0.918 = 91.8\%$

Hence, we can conclude that the Accuracy of Neural Network Model on our Dataset is around 91.8%

--END OF NEURAL NETWORK MODEL--

4. Discuss clustering algorithms that can be used for segmenting Champo Carpets's customers. Please justify your choices. Discuss what distance and similarity measures is suitable in this case.

Answer:

Clustering is a type of Unsupervised learning. Clustering algorithms are used to make small clusters for large dataset. When these clusters are made the items within clusters are like each other and close to each other in distance. The similarity should be high within each cluster and there should be low similarity between different clusters.

The different types of clustering methods are:

- Partition-based Algorithms (K-means algorithm)
- Hierarchy-based Algorithms

This distance can be calculated using various methods such as:

- Euclidian method
- Distance method

For our assignment we have used the Partition-based (K -means clustering) algorithm to segment the customers into various groups. It is the best choice when our dataset contains categorical variables. It is also less complex, fast, and highly scalable. We haven't used Hierarchical clustering as the as our data set does not contain variables that are hierarchical.

To calculate the distance, we have used Euclidian method over Manhattan as in Euclidian method the distance between any two objects is not dependent or affected by addition of any other new objects. When utilizing the Manhattan distance metric, the distortion is greater than when using the Euclidean distance metric. And it's well known that the optimal grouping of data-items is achieved with the least amount of distortion.

5. Discuss the data strategy for building customer segmentation using clustering. What are the benefits Champo Carpets can expect from clustering? Hint: Data strategy should clearly identify the data that should be used and how it should be used, including any feature engineering that may be performed before the model building.

Answer:

For customer segmentation we have implemented the K means clustering algorithm on the dataset. We used the sheet named "Data for clustering" under the Champo Carpets dataset as it consists of only unique records. Hence, we don't need to perform any data cleaning on it before processing it. To begin with we will first need to omit null values if any to begin with clustering. The next step would be to normalise the data. We can use the min-max function to do so. The next step would be to convert all the variables to numeric type which is a requirement of performing clustering. After this, we need to create clusters by using the 'kmean' function by specifying the dataset we are using, number of clusters we want to

create(using 'centres' parameter) and number of centroids we want to start with(using the 'nstart' parameter). The desired clusters can be now visualised using the 'fviz_cluster' function under the 'factoextra' library. The next step would be to make a scree plot for various values of k and use both 'WSS' and 'Silhouette' methods to evaluate the model. Once the scree plot is created for both evaluation measures, we can select the best value for k using the elbow method. Once we have decided the best value of k, we can create a model using that value for k to get the best possible results.

CHAMPO CARPETS: K-MEANS CLUSTERING

```

library(dplyr)

library(readxl)
library(tidyr)
library(ggplot2)

library(ISLR)
library(rpart)
library(rpart.plot)
library(psych)

library(ggplot2)
library(Rcpp)
library(fastDummies)

library(factoextra)

library(klaR)

library(purrr)
library(ggplot2)
library(gridExtra)

Clustering_data <- read_excel("Clustering_data.xlsx")
View(Clustering_data)
mydata <- Clustering_data
mydata

mydata$Row_Labels <- as.numeric(factor(as.matrix(mydata$Row_Labels)))
summary(mydata)

##      Row_Labels Sum of QtyRequired Sum of TotalArea      Sum of Amount
## Min.      : 1   Min.      :      2   Min.      :      1.35   Min.      :      329
## 1st Qu.:12    1st Qu.:   565   1st Qu.:   376.77   1st Qu.:   39701
## Median :23    Median :  1566   Median :  2120.00   Median :  116778
## Mean   :23    Mean   : 12978   Mean   : 13056.59   Mean   :  698210
## 3rd Qu.:34    3rd Qu.: 11146   3rd Qu.:  8451.56   3rd Qu.:  426626
## Max.    :45    Max.    :183206   Max.    :209725.22   Max.    :11341053
##      DURRY      HANDLOOM      DOUBLE BACK      JACQUARD
## Min.      :      0   Min.      :      0.0   Min.      :      0.0   Min.      :      0.00
## 1st Qu.:      0   1st Qu.:      0.0   1st Qu.:      0.0   1st Qu.:      0.00
## Median :    289   Median :      0.0   Median :      0.0   Median :      0.00
## Mean   :   7103   Mean   :   185.5   Mean   :   407.9   Mean   :   89.42
## 3rd Qu.:   1560   3rd Qu.:      0.0   3rd Qu.:   175.0   3rd Qu.:   72.00
## Max.    :139618   Max.    :3673.0   Max.    :5439.0   Max.    :714.00
##      HAND TUFTED      HAND WOVEN      KNOTTED      GUN TUFTED
## Min.      :      0   Min.      :      0.0   Min.      :      0.0   Min.      :      0.000
## 1st Qu.:      0   1st Qu.:      0.0   1st Qu.:      0.0   1st Qu.:      0.000

```

K-MEANS CLUSTERING

```
## Median : 510      Median : 0.0      Median : 0.0      Median : 0.000
## Mean   : 3651     Mean   : 867.7     Mean   : 365.8     Mean   : 8.133
## 3rd Qu.: 3544     3rd Qu.: 269.0     3rd Qu.: 18.0      3rd Qu.: 0.000
## Max.   :60685     Max.   :14314.0    Max.   :9502.0     Max.   :195.000
## Powerloom Jacquard INDO TEBETAN
## Min.    : 0.0      Min.    : 0.0000
## 1st Qu.: 0.0      1st Qu.: 0.0000
## Median : 0.0      Median : 0.0000
## Mean    : 216.7     Mean    : 0.7111
## 3rd Qu.: 0.0      3rd Qu.: 0.0000
## Max.    :9753.0     Max.    :20.0000

sum(is.na(mydata))

## [1] 0

mydata <- na.omit(mydata)

myscale <- function(x) {
  (x - min(x)) / (max(x) - min(x))
}
df <- mydata %>% mutate_if(is.numeric, myscale)

kmModel <- kmeans(df, centers = 2, nstart = 100)

kmModel

## K-means clustering with 2 clusters of sizes 3, 42
##
## Cluster means:
##   Row_Labels Sum of QtyRequired Sum of TotalArea Sum of Amount      DURRY
## 1 0.5757576      0.45163133      0.49095247      0.25954037 0.39638394
## 2 0.4945887      0.04362532      0.03162803      0.04739456 0.02619529
##   HANDLOOM DOUBLE BACK JACQUARD HAND TUFTED HAND WOVEN      KNOTTED
## 1 0.44432344 0.62082491 0.4495798 0.17390898 0.26605654 0.460534624
## 2 0.02238342 0.03600977 0.1020742 0.05203569 0.04594403 0.008346614
##   GUN TUFTED Powerloom Jacquard INDO TEBETAN
## 1 0.33333333      0.3333333      0.00000000
## 2 0.02087912      0.0000000      0.03809524
##
## Clustering vector:
## [1] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 1 2 2 2 2 2 2 2 2 1 2 2 2 1 2 2 2
## [39] 2 2 2 2 2 2 2
##
## Within cluster sum of squares by cluster:
## [1] 4.805835 12.093430
## (between_SS / total_SS = 21.8 %)
##
## Available components:
```

K-MEANS CLUSTERING

```
##
## [1] "cluster"      "centers"      "totss"      "withinss"    "tot.withi
nss"
## [6] "betweenss"    "size"        "iter"      "ifault"

model <- lm(df$Row_Labels ~ ., data = df)
cooksD <- cooks.distance(model)
influential <- cooksD[(cooksD > (3 * mean(cooksD, na.rm = TRUE)))]
influential

##      <NA>      31      32      44
##      NA 67.03154 14.44764 14.54626

kmModel$cluster

## [1] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 1 2 2 2 2 2 2 2 2 1 2 2 2 1 2 2 2
2 2 2
## [39] 2 2 2 2 2 2 2

kmModel$centers

##   Row_Labels Sum of QtyRequired Sum of TotalArea Sum of Amount      DURRY
## 1  0.5757576      0.45163133      0.49095247      0.25954037 0.39638394
## 2  0.4945887      0.04362532      0.03162803      0.04739456 0.02619529
##   HANDLOOM DOUBLE BACK  JACQUARD HAND TUFTED HAND WOVEN      KNOTTED
## 1  0.44432344  0.62082491 0.4495798  0.17390898 0.26605654 0.460534624
## 2  0.02238342  0.03600977 0.1020742  0.05203569 0.04594403 0.008346614
##   GUN TUFTED Powerloom Jacquard INDO TEBETAN
## 1  0.33333333      0.3333333  0.00000000
## 2  0.02087912      0.0000000  0.03809524

kmModel$withinss

## [1]  4.805835 12.093430

kmModel$tot.withinss

## [1] 16.89927

kmModel$betweenss

## [1] 4.717582

kmModel$size

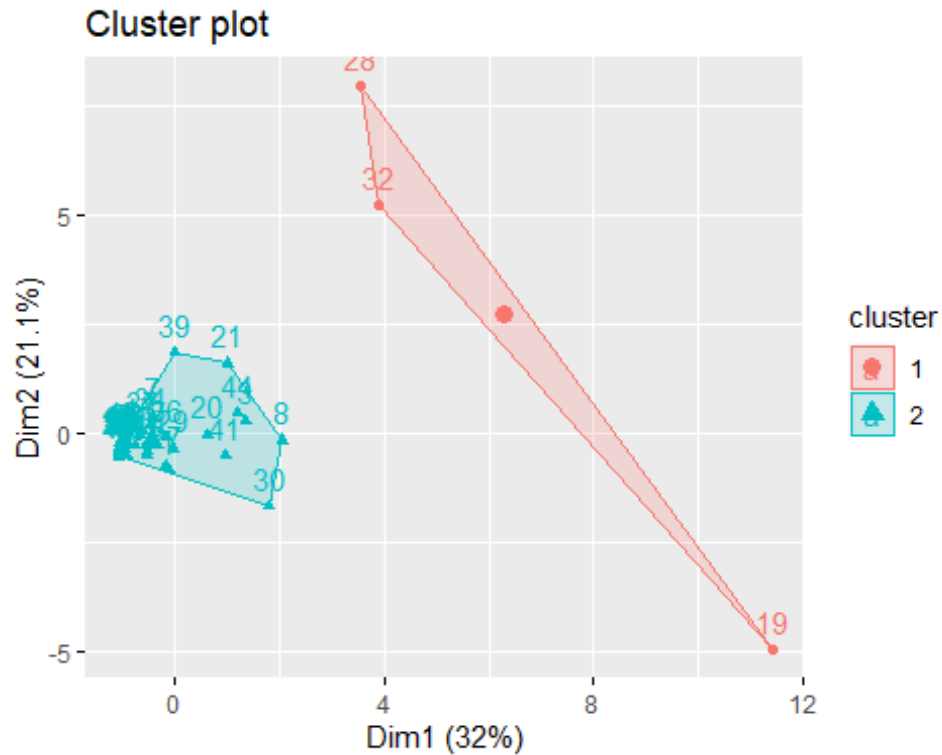
## [1]  3 42

kmModel$iter

## [1] 1

fviz_cluster(kmModel, data = df)
```

K-MEANS CLUSTERING

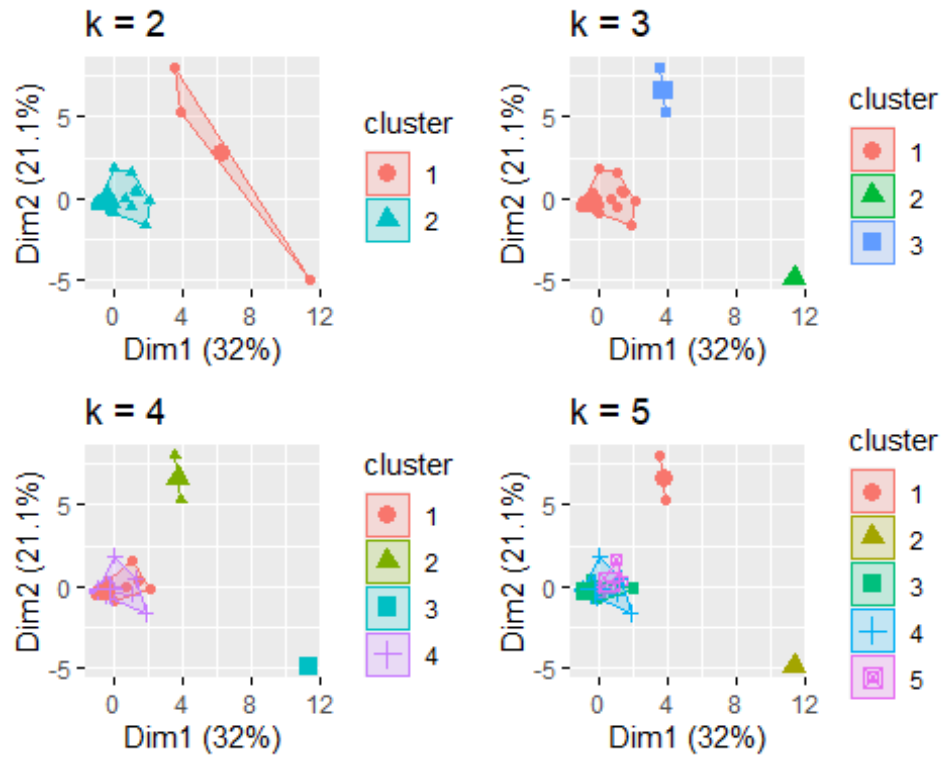


```
kmMode2 <- kmeans(df, centers = 3, nstart = 100)
kmMode3 <- kmeans(df, centers = 4, nstart = 100)
kmMode4 <- kmeans(df, centers = 5, nstart = 100)

fp1 <- fviz_cluster(kmModel, geom = "point", data = df) + ggtitle("k = 2")
fp2 <- fviz_cluster(kmMode2, geom = "point", data = df) + ggtitle("k = 3")
fp3 <- fviz_cluster(kmMode3, geom = "point", data = df) + ggtitle("k = 4")
fp4 <- fviz_cluster(kmMode4, geom = "point", data = df) + ggtitle("k = 5")

library(gridExtra)
grid.arrange(fp1, fp2, fp3, fp4, nrow = 2)
```

K-MEANS CLUSTERING

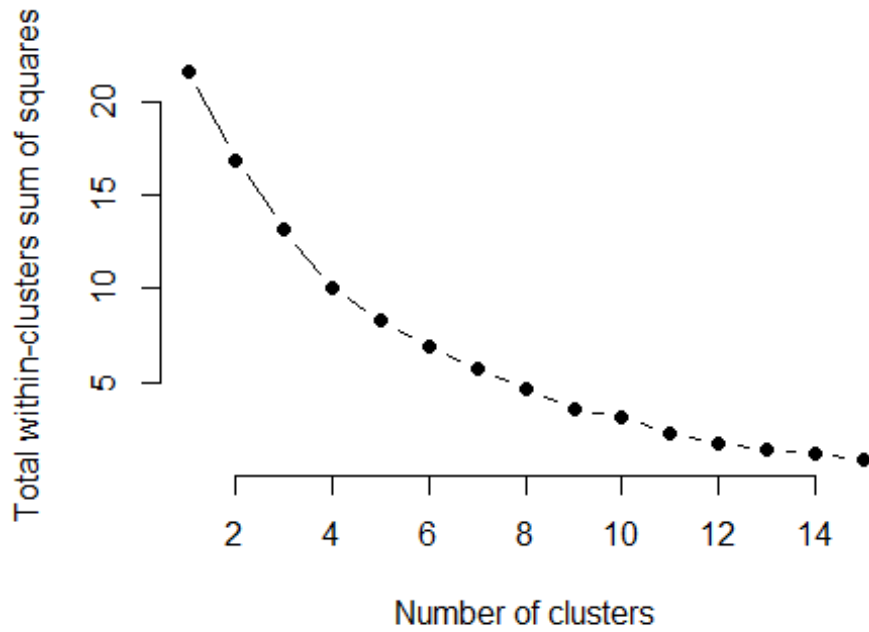


```
set.seed(123)
wss <- function(k) {
  kmeans(df, centers = k, nstart = 100)$tot.withinss
}

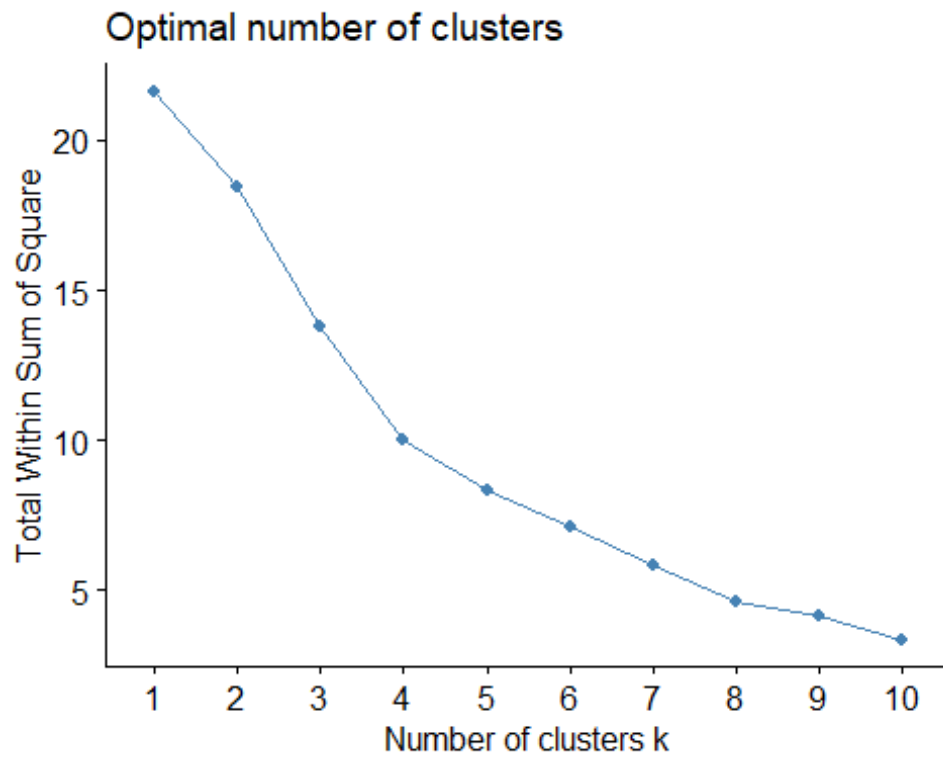
k.values <- 1:15
wss_values <- map_dbl(k.values, wss)

plot(k.values, wss_values,
     type="b", pch = 19, frame = FALSE,
     xlab="Number of clusters",
     ylab="Total within-clusters sum of squares")
```


K-MEANS CLUSTERING



```
set.seed(123)  
fviz_nbclust(df, kmeans, method = "wss")
```



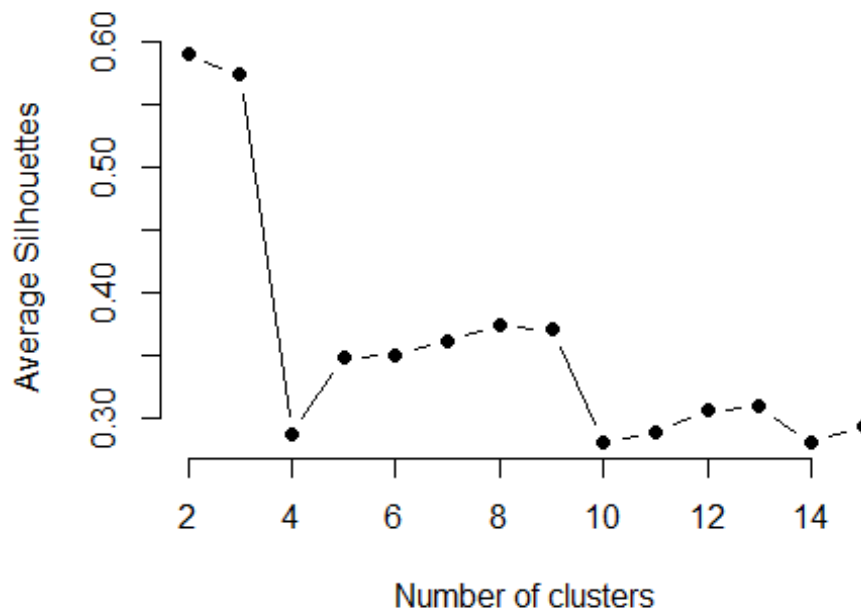
K-MEANS CLUSTERING

```
library(cluster)
avgsil <- function(k) {
  kmModel <- kmeans(df, centers = k, nstart = 100)
  ss <- silhouette(kmModel$cluster, dist(df))
  mean(ss[, 3])
}

k.values <- 2:15
avgsil_values <- map_dbl(k.values, avgsil)
map_dbl

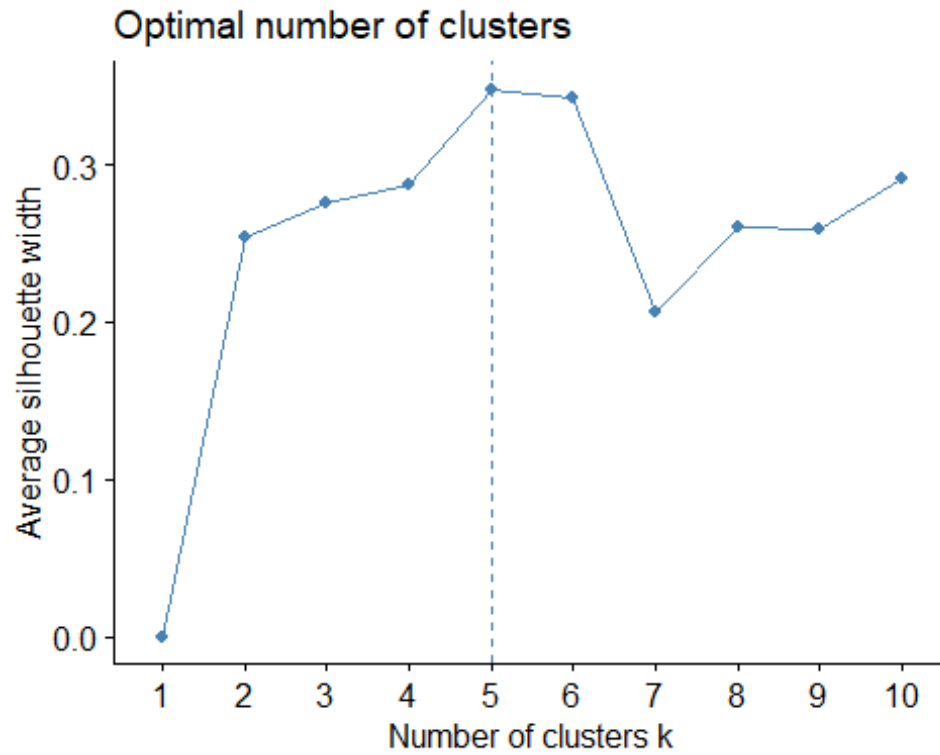
## function (.x, .f, ...)
## {
##   .f <- as_mapper(.f, ...)
##   .Call(map_impl, environment(), ".x", ".f", "double")
## }
## <bytecode: 0x000000002790ad90>
## <environment: namespace:purrr>

plot(k.values, avgsil_values,
     type = "b", pch = 19, frame = FALSE,
     xlab = "Number of clusters",
     ylab = "Average Silhouettes")
```

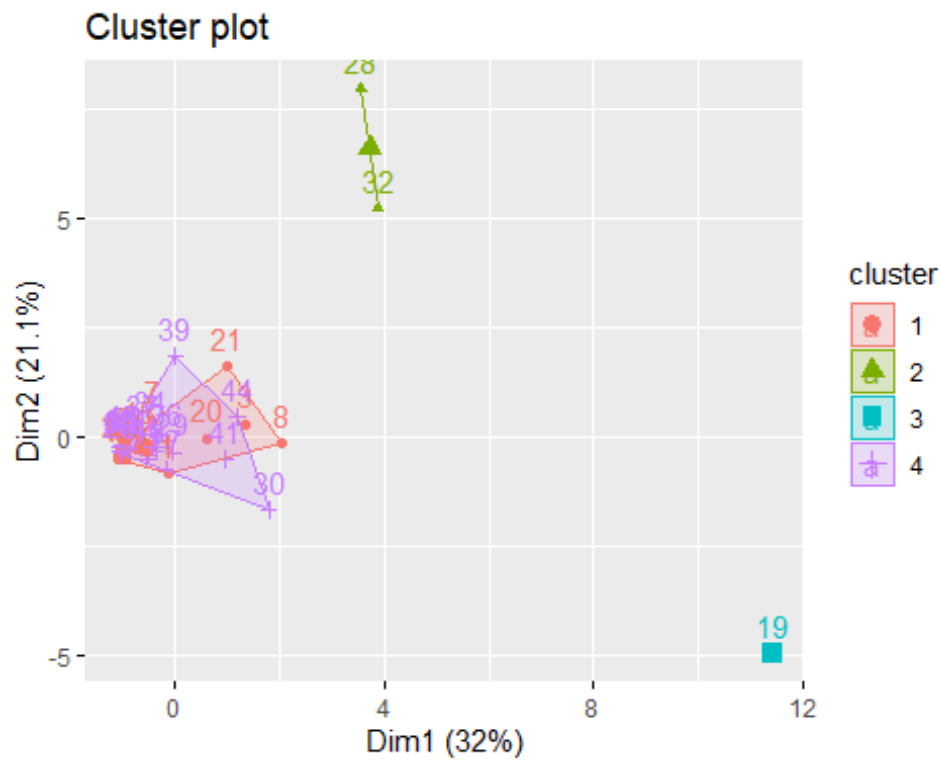


```
fviz_nbclust(df, kmeans, method = "silhouette")
```

K-MEANS CLUSTERING



```
fviz_cluster(kmMode3, data = df)
```



K-MEANS CLUSTERING

```
df %>%
  mutate(Cluster = kmMode3$cluster) %>%
  group_by(Cluster) %>%
  summarise_all("mean")

## # A tibble: 4 x 15
##   Cluster Row_Labels `Sum of QtyRequir~` `Sum of TotalAre~` `Sum of Amount`
DURRY
##   <int>         <dbl>         <dbl>         <dbl>         <dbl>
<dbl>
## 1         1         0.255         0.0291         0.0409         0.0245
0.0163
## 2         2         0.659         0.177         0.690         0.222
0.0946
## 3         3         0.409         1         0.0930         0.335
1
## 4         4         0.785         0.0612         0.0204         0.0751
0.0382
## # ... with 9 more variables: HANDLOOM <dbl>, DOUBLE BACK <dbl>, JACQUARD <
dbl>,
## #   HAND TUFTED <dbl>, HAND WOVEN <dbl>, KNOTTED <dbl>, GUN TUFTED <dbl>,
## #   Powerloom Jacquard <dbl>, INDO TEBETAN <dbl>

kmodes(df, modes = 4, iter.max = 10, weighted = FALSE)

## Warning in kmodes(df, modes = 4, iter.max = 10, weighted = FALSE): data ha
s
## numeric coloumns with more than 30 different levels!

## K-modes clustering with 4 clusters of sizes 15, 28, 1, 1
##
## Cluster modes:
##   Row_Labels Sum of QtyRequired Sum of TotalArea Sum of Amount      DURRY
## 1 0.13636364      0.003285954      0.002515319      0.004974449 0.000000000
## 2 0.00000000      0.000000000      0.000000000      0.000000000 0.000000000
## 3 0.29545455      0.006309906      0.013449351      0.010273535 0.002062771
## 4 0.04545455      0.103278313      0.255690036      0.140357085 0.025677205
##   HANDLOOM DOUBLE BACK JACQUARD HAND TUFTED HAND WOVEN   KNOTTED GUN TUFTE
D
## 1         0 0.00000000      0 0.002191645 0.00125751 0.00000000
0
## 2         0 0.00000000      0 0.000000000 0.00000000 0.00000000
0
## 3         0 0.01176687      0 0.013281701 0.00000000 0.00000000
0
## 4         0 0.03217503      1 0.193062536 0.14782730 0.0649337
0
##   Powerloom Jacquard INDO TEBETAN
## 1         0         0
## 2         0         0
## 3         0         0
```

K-MEANS CLUSTERING

```
## 4          0          0
##
## Clustering vector:
## [1] 2 2 4 2 2 2 1 1 2 2 2 2 2 3 2 2 2 1 1 1 1 2 2 2 2 2 2 1 1 2 1 1 2 2 2
1 1 1
## [39] 1 2 1 2 2 2 2
##
## Within cluster simple-matching distance by cluster:
## [1] 132 172  0  0
##
## Available components:
## [1] "cluster"  "size"      "modes"     "withindiff" "iterations"
## [6] "weighted"
```

RECOMMENDATION SYSTEM

CHAMPO CARPETS: RECOMMENDATION SYSTEM FOR CHAMPO CARPETS

```
RSD <- read_excel("C:/Users/Bodake/Desktop/UIC Sem 2/Data Mining IDS-572/R/RSD.xlsx")
mydata <- RSD
mydata <- as.data.frame(mydata)
sum(is.na(mydata))

## [1] 0

#str(mydata)
mydata[sapply(mydata, is.character)] <- lapply(mydata[sapply(mydata,
is.character)], as.factor)

df <- mydata[2:21]
str(mydata)

## 'data.frame':    20 obs. of  21 variables:
## $ Customer      : Factor w/ 20 levels "A-11","A-9","C-1",...: 6 14 10 2 4 8
12 20 3 18 ...
## $ HandTufted    : num  26612 2352 2697 11716 3816 ...
## $ DoubleWoven   : num  3000 5340 3085 2116 14314 ...
## $ Durry         : num  139618 25997 412 3585 37042 ...
## $ DoubleBack    : num  0 4691 5439 175 0 ...
## $ Knotted       : num  0 9502 3626 617 0 ...
## $ Jacquared     : num  550 353 60 714 0 231 0 100 0 0 ...
## $ Handloom      : num  3673 138 1085 0 0 ...
## $ Other         : num  9753 0 245 0 0 ...
## $ Rectangle     : num  132756 47951 16619 18923 55172 ...
## $ Square        : num  0 315 30 0 0 0 0 0 2 13 ...
## $ Round         : num  50540 107 0 0 0 ...
## $ Purple        : num  0 0 0 0 0 55 0 0 0 0 ...
## $ Gray          : num  0 1863 1216 273 2007 ...
## $ Navy          : num  16235 0 53 1843 1601 ...
## $ PINK          : num  0 0 138 552 0 316 0 875 0 0 ...
## $ BLUE          : num  15537 6218 69 705 0 ...
## $ BLUSHPINK     : num  0 0 0 0 0 0 0 250 0 0 ...
## $ NEUTRAL       : num  0 0 0 1832 0 ...
## $ TAN           : num  0 0 0 0 0 0 0 0 0 0 ...
## $ NAVY          : num  16235 0 53 1843 1601 ...

getCosine <- function(x,y)
{
  this.cosine <- sum(x*y) / (sqrt(sum(x*x)) * sqrt(sum(y*y)))
  return(this.cosine)
}
df.similarity <- matrix(NA,
nrow=ncol(df),ncol=ncol(df),dimnames=list(colnames(df),colnames(df)))
#df.similarity
```

RECOMMENDATION SYSTEM

```
for(i in 1:ncol(df)) {  
  for(j in 1:ncol(df)) {  
    df.similarity[i,j] <- getCosine(as.matrix(df[i]),as.matrix(df[j]))  
  }  
}  
  
df.similarity <- as.data.frame(df.similarity)  
  
df.neighbours <- matrix(NA,  
nrow=ncol(df.similarity),ncol=11,dimnames=list(colnames(df.similarity)))  
for(i in 1:ncol(df))  
{  
  df.neighbours[i,] <-  
(t(head(n=11,rownames(df.similarity[order(df.similarity[,i],decreasing=TRUE),  
][i]))))  
}  
df.neighbours[1:20]  
  
## [1] "HandTufted" "DoubleWoven" "Durry" "DoubleBack" "Knotted"  
## [6] "Jacquard" "Handloom" "Other" "Rectangle" "Square"  
## [11] "Round" "Purple" "Gray" "Navy" "PINK"  
## [16] "BLUE" "BLUSHPINK" "NEUTRAL" "TAN" "Navy"
```

Creating a Helper Function

```
getScore <- function(history, similarities)  
{  
  x <- sum(history*similarities)/sum(similarities)  
}
```

Creating a Matrix

```
holder <- matrix(NA,nrow=nrow(mydata),ncol=ncol(mydata)-  
1,dimnames=list((mydata$Customer),colnames(mydata[-1])))  
for(i in 1:nrow(holder))  
{  
  for(j in 1:ncol(holder))  
  {  
    Customer <- rownames(holder)[i]  
    product <- colnames(holder)[j]  
  
    if(as.integer(mydata[mydata$Customer==Customer,product]) == 1)  
    {  
      holder[i,j]<-" "  
    }  
  }  
}
```

RECOMMENDATION SYSTEM

```
} else {  
  topN<-  
  ((head(n=11,(df.similarity[order(df.similarity[,product],decreasing=TRUE),][product])))  
  topN.names <- as.character(rownames(topN))  
  topN.similarities <- as.numeric(topN[,1])  
  
  topN.similarities<-topN.similarities[-1]  
  topN.names<-topN.names[-1]  
  
  topN.purchases<- mydata[,c("Customer",topN.names)]  
  topN.CustomerPurchases<-  
  topN.purchases[topN.purchases$Customer==Customer,]  
  topN.CustomerPurchases <-  
  as.numeric(topN.CustomerPurchases[!(names(topN.CustomerPurchases) %in%  
  c("Customer"))])  
  
  holder[i,j]<-  
  getScore(similarities=topN.similarities,history=topN.CustomerPurchases)  
}  
}  
}  
mydata.Customer.scores <- holder  
  
topN<-  
((head(n=11,(df.similarity[order(df.similarity[,product],decreasing=TRUE),][product])))  
topN.names <- as.character(rownames(topN))  
topN.similarities <- as.numeric(topN[,1])  
  
topN.similarities<-topN.similarities[-1]  
topN.names<-topN.names[-1]  
  
topN.purchases<- mydata[,c("Customer",topN.names)]  
topN.CustomerPurchases<-topN.purchases[topN.purchases$Customer==Customer,]  
topN.CustomerPurchases <-  
as.numeric(topN.CustomerPurchases[!(names(topN.CustomerPurchases) %in%  
c("Customer"))])  
holder[i,j]<-  
getScore(similarities=topN.similarities,history=topN.CustomerPurchases)  
  
mydata.Customer.scores.holder <- matrix(NA,  
nrow=nrow(mydata.Customer.scores),ncol=100,dimnames=list(rownames(mydata.Customer.scores)))
```


RECOMMENDATION SYSTEM

```
for(i in 1:nrow(mydata.Customer.scores))
{
  mydata.Customer.scores.holder[i,] <-
names(head(n=100,(mydata.Customer.scores[,order(mydata.Customer.scores[i,],decreasing=TRUE)]))[i,]))
}
```

mydata.Customer.scores.holder

Following are the recommendations for the Consumers:

##	[,1]	[,2]	[,3]	[,4]	[,5]
## H-2	"Other"	"Handloom"	"BLUE"	"Navy"	"NAVY"
## P-5	"Navy"	"NAVY"	"Other"	"TAN"	"Round"
## M-1	"Rectangle"	"TAN"	"Square"	"DoubleWoven"	"Knotted"
## A-9	"TAN"	"BLUE"	"Navy"	"NAVY"	"Round"
## C-2	"Square"	"Durry"	"PINK"	"Knotted"	"Rectangle"
## JL	"TAN"	"DoubleWoven"	"Gray"	"Square"	"Navy"
## N-1	"PINK"	"Square"	"Knotted"	"Purple"	"DoubleBack"
## T-5	"BLUSHPINK"	"Jacquard"	"Gray"	"Durry"	"PINK"
## C-1	"Durry"	"Handloom"	"Round"	"DoubleWoven"	"Other"
## T-2	"DoubleWoven"	"Durry"	"Other"	"Handloom"	"Round"
## I-2	"Knotted"	"DoubleBack"	"Rectangle"	"TAN"	"BLUE"
## PD	"Purple"	"NEUTRAL"	"DoubleWoven"	"BLUE"	"Other"
## L-5	"DoubleWoven"	"Other"	"HandTufted"	"Round"	"Navy"
## M-2	"PINK"	"Square"	"Rectangle"	"Purple"	"Knotted"
## RC	"DoubleWoven"	"BLUE"	"Navy"	"NAVY"	"Square"
## P-4	"DoubleWoven"	"HandTufted"	"Round"	"Navy"	"NAVY"
## T-4	"Durry"	"HandTufted"	"BLUSHPINK"	"Jacquard"	"Gray"
## PC	"TAN"	"BLUE"	"Navy"	"NAVY"	"Durry"
## A-11	"Other"	"Round"	"BLUE"	"Navy"	"NAVY"
## CC	"Round"	"Handloom"	"Other"	"DoubleWoven"	"HandTufted"
##	[,6]	[,7]	[,8]	[,9]	[,10]
## H-2	"HandTufted"	"Round"	"DoubleWoven"	"Jacquard"	"TAN"
## P-5	"Handloom"	"BLUE"	"Jacquard"	"BLUSHPINK"	"Knotted"
## M-1	"Gray"	"Purple"	"BLUE"	"DoubleBack"	"Durry"
## A-9	"Handloom"	"Other"	"DoubleWoven"	"Durry"	"HandTufted"
## C-2	"DoubleBack"	"Purple"	"NEUTRAL"	"DoubleWoven"	"BLUE"
## JL	"NAVY"	"HandTufted"	"BLUE"	"Handloom"	"Knotted"
## N-1	"TAN"	"BLUE"	"Navy"	"NAVY"	"Other"
## T-5	"Square"	"Knotted"	"Rectangle"	"DoubleBack"	"NEUTRAL"
## C-1	"HandTufted"	"NEUTRAL"	"Gray"	"Square"	"BLUSHPINK"
## T-2	"HandTufted"	"Square"	"Knotted"	"Gray"	"Purple"
## I-2	"Navy"	"NAVY"	"Durry"	"DoubleWoven"	"Other"
## PD	"Round"	"Navy"	"NAVY"	"Handloom"	"HandTufted"
## L-5	"NAVY"	"Handloom"	"BLUE"	"BLUSHPINK"	"Gray"
## M-2	"DoubleBack"	"TAN"	"BLUE"	"Navy"	"NAVY"
## RC	"Handloom"	"Other"	"HandTufted"	"Round"	"Durry"

RECOMMENDATION SYSTEM

## P-4	"Handloom"	"BLUE"	"Other"	"Jacquared"	"TAN"
## T-4	"PINK"	"NEUTRAL"	"Square"	"Rectangle"	"Knotted"
## PC	"Round"	"Handloom"	"Other"	"DoubleWoven"	"NEUTRAL"
## A-11	"HandTufted"	"Jacquared"	"DoubleWoven"	"Durry"	"TAN"
## CC	"NEUTRAL"	"BLUSHPINK"	"Jacquared"	"Gray"	"Rectangle"
##	[,11]	[,12]	[,13]	[,14]	[,15]
## H-2	"Rectangle"	"Durry"	"Gray"	"BLUSHPINK"	"Square"
## P-5	"Durry"	"DoubleBack"	"PINK"	"Purple"	"Rectangle"
## M-1	"HandTufted"	"Navy"	"NAVY"	"Round"	"Other"
## A-9	"Jacquared"	"BLUSHPINK"	"NEUTRAL"	"Gray"	"PINK"
## C-2	"Other"	"HandTufted"	"Handloom"	"Navy"	"NAVY"
## JL	"Other"	"Round"	"Durry"	"BLUSHPINK"	"Jacquared"
## N-1	"Round"	"Handloom"	"Durry"	"HandTufted"	"DoubleWoven"
## T-5	"Purple"	"DoubleWoven"	"TAN"	"BLUE"	"Other"
## C-1	"PINK"	"Jacquared"	"Purple"	"DoubleBack"	"Rectangle"
## T-2	"Jacquared"	"PINK"	"BLUSHPINK"	"NEUTRAL"	"DoubleBack"
## I-2	"Round"	"HandTufted"	"Handloom"	"Gray"	"Jacquared"
## PD	"Gray"	"Jacquared"	"BLUSHPINK"	"TAN"	"Square"
## L-5	"Jacquared"	"TAN"	"Square"	"PINK"	"Knotted"
## M-2	"Handloom"	"Other"	"DoubleWoven"	"Round"	"Durry"
## RC	"Gray"	"NEUTRAL"	"BLUSHPINK"	"PINK"	"Purple"
## P-4	"BLUSHPINK"	"Gray"	"Durry"	"Square"	"Rectangle"
## T-4	"Purple"	"DoubleBack"	"TAN"	"BLUE"	"Navy"
## PC	"BLUSHPINK"	"PINK"	"HandTufted"	"Gray"	"Jacquared"
## A-11	"Handloom"	"Gray"	"BLUSHPINK"	"Knotted"	"Square"
## CC	"Purple"	"Knotted"	"DoubleBack"	"TAN"	"BLUE"
##	[,16]	[,17]	[,18]	[,19]	[,20]
## H-2	"PINK"	"Knotted"	"NEUTRAL"	"DoubleBack"	"Purple"
## P-5	"NEUTRAL"	"DoubleWoven"	"Square"	"Gray"	"HandTufted"
## M-1	"PINK"	"BLUSHPINK"	"Handloom"	"Jacquared"	"NEUTRAL"
## A-9	"Square"	"Rectangle"	"Purple"	"Knotted"	"DoubleBack"
## C-2	"Round"	"Gray"	"TAN"	"Jacquared"	"BLUSHPINK"
## JL	"Purple"	"PINK"	"DoubleBack"	"NEUTRAL"	"Rectangle"
## N-1	"Jacquared"	"Rectangle"	"NEUTRAL"	"BLUSHPINK"	"Gray"
## T-5	"Round"	"Handloom"	"Navy"	"NAVY"	"HandTufted"
## C-1	"Knotted"	"TAN"	"BLUE"	"Navy"	"NAVY"
## T-2	"Rectangle"	"TAN"	"BLUE"	"Navy"	"NAVY"
## I-2	"BLUSHPINK"	"NEUTRAL"	"Square"	"PINK"	"Purple"
## PD	"PINK"	"Knotted"	"Durry"	"Rectangle"	"DoubleBack"
## L-5	"Durry"	"Rectangle"	"DoubleBack"	"Purple"	"NEUTRAL"
## M-2	"HandTufted"	"BLUSHPINK"	"Gray"	"NEUTRAL"	"Jacquared"
## RC	"Jacquared"	"Knotted"	"DoubleBack"	"Rectangle"	"TAN"
## P-4	"Knotted"	"PINK"	"DoubleBack"	"Purple"	"NEUTRAL"
## T-4	"NAVY"	"Round"	"Other"	"DoubleWoven"	"Handloom"
## PC	"Purple"	"Rectangle"	"Knotted"	"DoubleBack"	"Square"
## A-11	"Purple"	"DoubleBack"	"Rectangle"	"PINK"	"NEUTRAL"
## CC	"Navy"	"NAVY"	"Durry"	"Square"	"PINK"
##	[,21]	[,22]	[,23]	[,24]	[,25]
## H-2	"Other"	"Handloom"	"BLUE"	"Navy"	"NAVY"
## P-5	"Navy"	"NAVY"	"Other"	"TAN"	"Round"

RECOMMENDATION SYSTEM

## M-1	"Rectangle"	"TAN"	"Square"	"DoubleWoven"	"Knotted"
## A-9	"TAN"	"BLUE"	"Navy"	"NAVY"	"Round"
## C-2	"Square"	"Durry"	"PINK"	"Knotted"	"Rectangle"
## JL	"TAN"	"DoubleWoven"	"Gray"	"Square"	"Navy"
## N-1	"PINK"	"Square"	"Knotted"	"Purple"	"DoubleBack"
## T-5	"BLUSHPINK"	"Jacquard"	"Gray"	"Durry"	"PINK"
## C-1	"Durry"	"Handloom"	"Round"	"DoubleWoven"	"Other"
## T-2	"DoubleWoven"	"Durry"	"Other"	"Handloom"	"Round"
## I-2	"Knotted"	"DoubleBack"	"Rectangle"	"TAN"	"BLUE"
## PD	"Purple"	"NEUTRAL"	"DoubleWoven"	"BLUE"	"Other"
## L-5	"DoubleWoven"	"Other"	"HandTufted"	"Round"	"Navy"
## M-2	"PINK"	"Square"	"Rectangle"	"Purple"	"Knotted"
## RC	"DoubleWoven"	"BLUE"	"Navy"	"NAVY"	"Square"
## P-4	"DoubleWoven"	"HandTufted"	"Round"	"Navy"	"NAVY"
## T-4	"Durry"	"HandTufted"	"BLUSHPINK"	"Jacquard"	"Gray"
## PC	"TAN"	"BLUE"	"Navy"	"NAVY"	"Durry"
## A-11	"Other"	"Round"	"BLUE"	"Navy"	"NAVY"
## CC	"Round"	"Handloom"	"Other"	"DoubleWoven"	"HandTufted"
##	[,26]	[,27]	[,28]	[,29]	[,30]
## H-2	"HandTufted"	"Round"	"DoubleWoven"	"Jacquard"	"TAN"
## P-5	"Handloom"	"BLUE"	"Jacquard"	"BLUSHPINK"	"Knotted"
## M-1	"Gray"	"Purple"	"BLUE"	"DoubleBack"	"Durry"
## A-9	"Handloom"	"Other"	"DoubleWoven"	"Durry"	"HandTufted"
## C-2	"DoubleBack"	"Purple"	"NEUTRAL"	"DoubleWoven"	"BLUE"
## JL	"NAVY"	"HandTufted"	"BLUE"	"Handloom"	"Knotted"
## N-1	"TAN"	"BLUE"	"Navy"	"NAVY"	"Other"
## T-5	"Square"	"Knotted"	"Rectangle"	"DoubleBack"	"NEUTRAL"
## C-1	"HandTufted"	"NEUTRAL"	"Gray"	"Square"	"BLUSHPINK"
## T-2	"HandTufted"	"Square"	"Knotted"	"Gray"	"Purple"
## I-2	"Navy"	"NAVY"	"Durry"	"DoubleWoven"	"Other"
## PD	"Round"	"Navy"	"NAVY"	"Handloom"	"HandTufted"
## L-5	"NAVY"	"Handloom"	"BLUE"	"BLUSHPINK"	"Gray"
## M-2	"DoubleBack"	"TAN"	"BLUE"	"Navy"	"NAVY"
## RC	"Handloom"	"Other"	"HandTufted"	"Round"	"Durry"
## P-4	"Handloom"	"BLUE"	"Other"	"Jacquard"	"TAN"
## T-4	"PINK"	"NEUTRAL"	"Square"	"Rectangle"	"Knotted"
## PC	"Round"	"Handloom"	"Other"	"DoubleWoven"	"NEUTRAL"
## A-11	"HandTufted"	"Jacquard"	"DoubleWoven"	"Durry"	"TAN"
## CC	"NEUTRAL"	"BLUSHPINK"	"Jacquard"	"Gray"	"Rectangle"
##	[,31]	[,32]	[,33]	[,34]	[,35]
## H-2	"Rectangle"	"Durry"	"Gray"	"BLUSHPINK"	"Square"
## P-5	"Durry"	"DoubleBack"	"PINK"	"Purple"	"Rectangle"
## M-1	"HandTufted"	"Navy"	"NAVY"	"Round"	"Other"
## A-9	"Jacquard"	"BLUSHPINK"	"NEUTRAL"	"Gray"	"PINK"
## C-2	"Other"	"HandTufted"	"Handloom"	"Navy"	"NAVY"
## JL	"Other"	"Round"	"Durry"	"BLUSHPINK"	"Jacquard"
## N-1	"Round"	"Handloom"	"Durry"	"HandTufted"	"DoubleWoven"
## T-5	"Purple"	"DoubleWoven"	"TAN"	"BLUE"	"Other"
## C-1	"PINK"	"Jacquard"	"Purple"	"DoubleBack"	"Rectangle"
## T-2	"Jacquard"	"PINK"	"BLUSHPINK"	"NEUTRAL"	"DoubleBack"

RECOMMENDATION SYSTEM

## I-2	"Round"	"HandTufted"	"Handloom"	"Gray"	"Jacquared"
## PD	"Gray"	"Jacquared"	"BLUSHPINK"	"TAN"	"Square"
## L-5	"Jacquared"	"TAN"	"Square"	"PINK"	"Knotted"
## M-2	"Handloom"	"Other"	"DoubleWoven"	"Round"	"Durry"
## RC	"Gray"	"NEUTRAL"	"BLUSHPINK"	"PINK"	"Purple"
## P-4	"BLUSHPINK"	"Gray"	"Durry"	"Square"	"Rectangle"
## T-4	"Purple"	"DoubleBack"	"TAN"	"BLUE"	"Navy"
## PC	"BLUSHPINK"	"PINK"	"HandTufted"	"Gray"	"Jacquared"
## A-11	"Handloom"	"Gray"	"BLUSHPINK"	"Knotted"	"Square"
## CC	"Purple"	"Knotted"	"DoubleBack"	"TAN"	"BLUE"
##	[,36]	[,37]	[,38]	[,39]	[,40]
## H-2	"PINK"	"Knotted"	"NEUTRAL"	"DoubleBack"	"Purple"
## P-5	"NEUTRAL"	"DoubleWoven"	"Square"	"Gray"	"HandTufted"
## M-1	"PINK"	"BLUSHPINK"	"Handloom"	"Jacquared"	"NEUTRAL"
## A-9	"Square"	"Rectangle"	"Purple"	"Knotted"	"DoubleBack"
## C-2	"Round"	"Gray"	"TAN"	"Jacquared"	"BLUSHPINK"
## JL	"Purple"	"PINK"	"DoubleBack"	"NEUTRAL"	"Rectangle"
## N-1	"Jacquared"	"Rectangle"	"NEUTRAL"	"BLUSHPINK"	"Gray"
## T-5	"Round"	"Handloom"	"Navy"	"NAVY"	"HandTufted"
## C-1	"Knotted"	"TAN"	"BLUE"	"Navy"	"NAVY"
## T-2	"Rectangle"	"TAN"	"BLUE"	"Navy"	"NAVY"
## I-2	"BLUSHPINK"	"NEUTRAL"	"Square"	"PINK"	"Purple"
## PD	"PINK"	"Knotted"	"Durry"	"Rectangle"	"DoubleBack"
## L-5	"Durry"	"Rectangle"	"DoubleBack"	"Purple"	"NEUTRAL"
## M-2	"HandTufted"	"BLUSHPINK"	"Gray"	"NEUTRAL"	"Jacquared"
## RC	"Jacquared"	"Knotted"	"DoubleBack"	"Rectangle"	"TAN"
## P-4	"Knotted"	"PINK"	"DoubleBack"	"Purple"	"NEUTRAL"
## T-4	"NAVY"	"Round"	"Other"	"DoubleWoven"	"Handloom"
## PC	"Purple"	"Rectangle"	"Knotted"	"DoubleBack"	"Square"
## A-11	"Purple"	"DoubleBack"	"Rectangle"	"PINK"	"NEUTRAL"
## CC	"Navy"	"NAVY"	"Durry"	"Square"	"PINK"
##	[,41]	[,42]	[,43]	[,44]	[,45]
## H-2	"Other"	"Handloom"	"BLUE"	"Navy"	"NAVY"
## P-5	"Navy"	"NAVY"	"Other"	"TAN"	"Round"
## M-1	"Rectangle"	"TAN"	"Square"	"DoubleWoven"	"Knotted"
## A-9	"TAN"	"BLUE"	"Navy"	"NAVY"	"Round"
## C-2	"Square"	"Durry"	"PINK"	"Knotted"	"Rectangle"
## JL	"TAN"	"DoubleWoven"	"Gray"	"Square"	"Navy"
## N-1	"PINK"	"Square"	"Knotted"	"Purple"	"DoubleBack"
## T-5	"BLUSHPINK"	"Jacquared"	"Gray"	"Durry"	"PINK"
## C-1	"Durry"	"Handloom"	"Round"	"DoubleWoven"	"Other"
## T-2	"DoubleWoven"	"Durry"	"Other"	"Handloom"	"Round"
## I-2	"Knotted"	"DoubleBack"	"Rectangle"	"TAN"	"BLUE"
## PD	"Purple"	"NEUTRAL"	"DoubleWoven"	"BLUE"	"Other"
## L-5	"DoubleWoven"	"Other"	"HandTufted"	"Round"	"Navy"
## M-2	"PINK"	"Square"	"Rectangle"	"Purple"	"Knotted"
## RC	"DoubleWoven"	"BLUE"	"Navy"	"NAVY"	"Square"
## P-4	"DoubleWoven"	"HandTufted"	"Round"	"Navy"	"NAVY"
## T-4	"Durry"	"HandTufted"	"BLUSHPINK"	"Jacquared"	"Gray"
## PC	"TAN"	"BLUE"	"Navy"	"NAVY"	"Durry"

RECOMMENDATION SYSTEM

## A-11	"Other"	"Round"	"BLUE"	"Navy"	"NAVY"
## CC	"Round"	"Handloom"	"Other"	"DoubleWoven"	"HandTufted"
##	[,46]	[,47]	[,48]	[,49]	[,50]
## H-2	"HandTufted"	"Round"	"DoubleWoven"	"Jacquared"	"TAN"
## P-5	"Handloom"	"BLUE"	"Jacquared"	"BLUSHPINK"	"Knotted"
## M-1	"Gray"	"Purple"	"BLUE"	"DoubleBack"	"Durry"
## A-9	"Handloom"	"Other"	"DoubleWoven"	"Durry"	"HandTufted"
## C-2	"DoubleBack"	"Purple"	"NEUTRAL"	"DoubleWoven"	"BLUE"
## JL	"NAVY"	"HandTufted"	"BLUE"	"Handloom"	"Knotted"
## N-1	"TAN"	"BLUE"	"Navy"	"NAVY"	"Other"
## T-5	"Square"	"Knotted"	"Rectangle"	"DoubleBack"	"NEUTRAL"
## C-1	"HandTufted"	"NEUTRAL"	"Gray"	"Square"	"BLUSHPINK"
## T-2	"HandTufted"	"Square"	"Knotted"	"Gray"	"Purple"
## I-2	"Navy"	"NAVY"	"Durry"	"DoubleWoven"	"Other"
## PD	"Round"	"Navy"	"NAVY"	"Handloom"	"HandTufted"
## L-5	"NAVY"	"Handloom"	"BLUE"	"BLUSHPINK"	"Gray"
## M-2	"DoubleBack"	"TAN"	"BLUE"	"Navy"	"NAVY"
## RC	"Handloom"	"Other"	"HandTufted"	"Round"	"Durry"
## P-4	"Handloom"	"BLUE"	"Other"	"Jacquared"	"TAN"
## T-4	"PINK"	"NEUTRAL"	"Square"	"Rectangle"	"Knotted"
## PC	"Round"	"Handloom"	"Other"	"DoubleWoven"	"NEUTRAL"
## A-11	"HandTufted"	"Jacquared"	"DoubleWoven"	"Durry"	"TAN"
## CC	"NEUTRAL"	"BLUSHPINK"	"Jacquared"	"Gray"	"Rectangle"
##	[,51]	[,52]	[,53]	[,54]	[,55]
## H-2	"Rectangle"	"Durry"	"Gray"	"BLUSHPINK"	"Square"
## P-5	"Durry"	"DoubleBack"	"PINK"	"Purple"	"Rectangle"
## M-1	"HandTufted"	"Navy"	"NAVY"	"Round"	"Other"
## A-9	"Jacquared"	"BLUSHPINK"	"NEUTRAL"	"Gray"	"PINK"
## C-2	"Other"	"HandTufted"	"Handloom"	"Navy"	"NAVY"
## JL	"Other"	"Round"	"Durry"	"BLUSHPINK"	"Jacquared"
## N-1	"Round"	"Handloom"	"Durry"	"HandTufted"	"DoubleWoven"
## T-5	"Purple"	"DoubleWoven"	"TAN"	"BLUE"	"Other"
## C-1	"PINK"	"Jacquared"	"Purple"	"DoubleBack"	"Rectangle"
## T-2	"Jacquared"	"PINK"	"BLUSHPINK"	"NEUTRAL"	"DoubleBack"
## I-2	"Round"	"HandTufted"	"Handloom"	"Gray"	"Jacquared"
## PD	"Gray"	"Jacquared"	"BLUSHPINK"	"TAN"	"Square"
## L-5	"Jacquared"	"TAN"	"Square"	"PINK"	"Knotted"
## M-2	"Handloom"	"Other"	"DoubleWoven"	"Round"	"Durry"
## RC	"Gray"	"NEUTRAL"	"BLUSHPINK"	"PINK"	"Purple"
## P-4	"BLUSHPINK"	"Gray"	"Durry"	"Square"	"Rectangle"
## T-4	"Purple"	"DoubleBack"	"TAN"	"BLUE"	"Navy"
## PC	"BLUSHPINK"	"PINK"	"HandTufted"	"Gray"	"Jacquared"
## A-11	"Handloom"	"Gray"	"BLUSHPINK"	"Knotted"	"Square"
## CC	"Purple"	"Knotted"	"DoubleBack"	"TAN"	"BLUE"
##	[,56]	[,57]	[,58]	[,59]	[,60]
## H-2	"PINK"	"Knotted"	"NEUTRAL"	"DoubleBack"	"Purple"
## P-5	"NEUTRAL"	"DoubleWoven"	"Square"	"Gray"	"HandTufted"
## M-1	"PINK"	"BLUSHPINK"	"Handloom"	"Jacquared"	"NEUTRAL"
## A-9	"Square"	"Rectangle"	"Purple"	"Knotted"	"DoubleBack"
## C-2	"Round"	"Gray"	"TAN"	"Jacquared"	"BLUSHPINK"

RECOMMENDATION SYSTEM

## JL	"Purple"	"PINK"	"DoubleBack"	"NEUTRAL "	"Rectangle"
## N-1	"Jacquared"	"Rectangle"	"NEUTRAL "	"BLUSHPINK"	"Gray"
## T-5	"Round"	"Handloom"	"Navy"	"NAVY"	"HandTufted"
## C-1	"Knotted"	"TAN"	"BLUE"	"Navy"	"NAVY"
## T-2	"Rectangle"	"TAN"	"BLUE"	"Navy"	"NAVY"
## I-2	"BLUSHPINK"	"NEUTRAL "	"Square"	"PINK"	"Purple"
## PD	"PINK"	"Knotted"	"Durry"	"Rectangle"	"DoubleBack"
## L-5	"Durry"	"Rectangle"	"DoubleBack"	"Purple"	"NEUTRAL "
## M-2	"HandTufted"	"BLUSHPINK"	"Gray"	"NEUTRAL "	"Jacquared"
## RC	"Jacquared"	"Knotted"	"DoubleBack"	"Rectangle"	"TAN"
## P-4	"Knotted"	"PINK"	"DoubleBack"	"Purple"	"NEUTRAL "
## T-4	"NAVY"	"Round"	"Other"	"DoubleWoven"	"Handloom"
## PC	"Purple"	"Rectangle"	"Knotted"	"DoubleBack"	"Square"
## A-11	"Purple"	"DoubleBack"	"Rectangle"	"PINK"	"NEUTRAL "
## CC	"Navy"	"NAVY"	"Durry"	"Square"	"PINK"
##	[,61]	[,62]	[,63]	[,64]	[,65]
## H-2	"Other"	"Handloom"	"BLUE"	"Navy"	"NAVY"
## P-5	"Navy"	"NAVY"	"Other"	"TAN"	"Round"
## M-1	"Rectangle"	"TAN"	"Square"	"DoubleWoven"	"Knotted"
## A-9	"TAN"	"BLUE"	"Navy"	"NAVY"	"Round"
## C-2	"Square"	"Durry"	"PINK"	"Knotted"	"Rectangle"
## JL	"TAN"	"DoubleWoven"	"Gray"	"Square"	"Navy"
## N-1	"PINK"	"Square"	"Knotted"	"Purple"	"DoubleBack"
## T-5	"BLUSHPINK"	"Jacquared"	"Gray"	"Durry"	"PINK"
## C-1	"Durry"	"Handloom"	"Round"	"DoubleWoven"	"Other"
## T-2	"DoubleWoven"	"Durry"	"Other"	"Handloom"	"Round"
## I-2	"Knotted"	"DoubleBack"	"Rectangle"	"TAN"	"BLUE"
## PD	"Purple"	"NEUTRAL "	"DoubleWoven"	"BLUE"	"Other"
## L-5	"DoubleWoven"	"Other"	"HandTufted"	"Round"	"Navy"
## M-2	"PINK"	"Square"	"Rectangle"	"Purple"	"Knotted"
## RC	"DoubleWoven"	"BLUE"	"Navy"	"NAVY"	"Square"
## P-4	"DoubleWoven"	"HandTufted"	"Round"	"Navy"	"NAVY"
## T-4	"Durry"	"HandTufted"	"BLUSHPINK"	"Jacquared"	"Gray"
## PC	"TAN"	"BLUE"	"Navy"	"NAVY"	"Durry"
## A-11	"Other"	"Round"	"BLUE"	"Navy"	"NAVY"
## CC	"Round"	"Handloom"	"Other"	"DoubleWoven"	"HandTufted"
##	[,66]	[,67]	[,68]	[,69]	[,70]
## H-2	"HandTufted"	"Round"	"DoubleWoven"	"Jacquared"	"TAN"
## P-5	"Handloom"	"BLUE"	"Jacquared"	"BLUSHPINK"	"Knotted"
## M-1	"Gray"	"Purple"	"BLUE"	"DoubleBack"	"Durry"
## A-9	"Handloom"	"Other"	"DoubleWoven"	"Durry"	"HandTufted"
## C-2	"DoubleBack"	"Purple"	"NEUTRAL "	"DoubleWoven"	"BLUE"
## JL	"NAVY"	"HandTufted"	"BLUE"	"Handloom"	"Knotted"
## N-1	"TAN"	"BLUE"	"Navy"	"NAVY"	"Other"
## T-5	"Square"	"Knotted"	"Rectangle"	"DoubleBack"	"NEUTRAL "
## C-1	"HandTufted"	"NEUTRAL "	"Gray"	"Square"	"BLUSHPINK"
## T-2	"HandTufted"	"Square"	"Knotted"	"Gray"	"Purple"
## I-2	"Navy"	"NAVY"	"Durry"	"DoubleWoven"	"Other"
## PD	"Round"	"Navy"	"NAVY"	"Handloom"	"HandTufted"
## L-5	"NAVY"	"Handloom"	"BLUE"	"BLUSHPINK"	"Gray"

RECOMMENDATION SYSTEM

##	M-2	"DoubleBack"	"TAN"	"BLUE"	"Navy"	"NAVY"
##	RC	"Handloom"	"Other"	"HandTufted"	"Round"	"Durry"
##	P-4	"Handloom"	"BLUE"	"Other"	"Jacquared"	"TAN"
##	T-4	"PINK"	"NEUTRAL"	"Square"	"Rectangle"	"Knotted"
##	PC	"Round"	"Handloom"	"Other"	"DoubleWoven"	"NEUTRAL"
##	A-11	"HandTufted"	"Jacquared"	"DoubleWoven"	"Durry"	"TAN"
##	CC	"NEUTRAL"	"BLUSHPINK"	"Jacquared"	"Gray"	"Rectangle"
##		[,71]	[,72]	[,73]	[,74]	[,75]
##	H-2	"Rectangle"	"Durry"	"Gray"	"BLUSHPINK"	"Square"
##	P-5	"Durry"	"DoubleBack"	"PINK"	"Purple"	"Rectangle"
##	M-1	"HandTufted"	"Navy"	"NAVY"	"Round"	"Other"
##	A-9	"Jacquared"	"BLUSHPINK"	"NEUTRAL"	"Gray"	"PINK"
##	C-2	"Other"	"HandTufted"	"Handloom"	"Navy"	"NAVY"
##	JL	"Other"	"Round"	"Durry"	"BLUSHPINK"	"Jacquared"
##	N-1	"Round"	"Handloom"	"Durry"	"HandTufted"	"DoubleWoven"
##	T-5	"Purple"	"DoubleWoven"	"TAN"	"BLUE"	"Other"
##	C-1	"PINK"	"Jacquared"	"Purple"	"DoubleBack"	"Rectangle"
##	T-2	"Jacquared"	"PINK"	"BLUSHPINK"	"NEUTRAL"	"DoubleBack"
##	I-2	"Round"	"HandTufted"	"Handloom"	"Gray"	"Jacquared"
##	PD	"Gray"	"Jacquared"	"BLUSHPINK"	"TAN"	"Square"
##	L-5	"Jacquared"	"TAN"	"Square"	"PINK"	"Knotted"
##	M-2	"Handloom"	"Other"	"DoubleWoven"	"Round"	"Durry"
##	RC	"Gray"	"NEUTRAL"	"BLUSHPINK"	"PINK"	"Purple"
##	P-4	"BLUSHPINK"	"Gray"	"Durry"	"Square"	"Rectangle"
##	T-4	"Purple"	"DoubleBack"	"TAN"	"BLUE"	"Navy"
##	PC	"BLUSHPINK"	"PINK"	"HandTufted"	"Gray"	"Jacquared"
##	A-11	"Handloom"	"Gray"	"BLUSHPINK"	"Knotted"	"Square"
##	CC	"Purple"	"Knotted"	"DoubleBack"	"TAN"	"BLUE"
##		[,76]	[,77]	[,78]	[,79]	[,80]
##	H-2	"PINK"	"Knotted"	"NEUTRAL"	"DoubleBack"	"Purple"
##	P-5	"NEUTRAL"	"DoubleWoven"	"Square"	"Gray"	"HandTufted"
##	M-1	"PINK"	"BLUSHPINK"	"Handloom"	"Jacquared"	"NEUTRAL"
##	A-9	"Square"	"Rectangle"	"Purple"	"Knotted"	"DoubleBack"
##	C-2	"Round"	"Gray"	"TAN"	"Jacquared"	"BLUSHPINK"
##	JL	"Purple"	"PINK"	"DoubleBack"	"NEUTRAL"	"Rectangle"
##	N-1	"Jacquared"	"Rectangle"	"NEUTRAL"	"BLUSHPINK"	"Gray"
##	T-5	"Round"	"Handloom"	"Navy"	"NAVY"	"HandTufted"
##	C-1	"Knotted"	"TAN"	"BLUE"	"Navy"	"NAVY"
##	T-2	"Rectangle"	"TAN"	"BLUE"	"Navy"	"NAVY"
##	I-2	"BLUSHPINK"	"NEUTRAL"	"Square"	"PINK"	"Purple"
##	PD	"PINK"	"Knotted"	"Durry"	"Rectangle"	"DoubleBack"
##	L-5	"Durry"	"Rectangle"	"DoubleBack"	"Purple"	"NEUTRAL"
##	M-2	"HandTufted"	"BLUSHPINK"	"Gray"	"NEUTRAL"	"Jacquared"
##	RC	"Jacquared"	"Knotted"	"DoubleBack"	"Rectangle"	"TAN"
##	P-4	"Knotted"	"PINK"	"DoubleBack"	"Purple"	"NEUTRAL"
##	T-4	"NAVY"	"Round"	"Other"	"DoubleWoven"	"Handloom"
##	PC	"Purple"	"Rectangle"	"Knotted"	"DoubleBack"	"Square"
##	A-11	"Purple"	"DoubleBack"	"Rectangle"	"PINK"	"NEUTRAL"
##	CC	"Navy"	"NAVY"	"Durry"	"Square"	"PINK"
##		[,81]	[,82]	[,83]	[,84]	[,85]

RECOMMENDATION SYSTEM

##	H-2	"Other"	"Handloom"	"BLUE"	"Navy"	"NAVY"
##	P-5	"Navy"	"NAVY"	"Other"	"TAN"	"Round"
##	M-1	"Rectangle"	"TAN"	"Square"	"DoubleWoven"	"Knotted"
##	A-9	"TAN"	"BLUE"	"Navy"	"NAVY"	"Round"
##	C-2	"Square"	"Durry"	"PINK"	"Knotted"	"Rectangle"
##	JL	"TAN"	"DoubleWoven"	"Gray"	"Square"	"Navy"
##	N-1	"PINK"	"Square"	"Knotted"	"Purple"	"DoubleBack"
##	T-5	"BLUSHPINK"	"Jacquard"	"Gray"	"Durry"	"PINK"
##	C-1	"Durry"	"Handloom"	"Round"	"DoubleWoven"	"Other"
##	T-2	"DoubleWoven"	"Durry"	"Other"	"Handloom"	"Round"
##	I-2	"Knotted"	"DoubleBack"	"Rectangle"	"TAN"	"BLUE"
##	PD	"Purple"	"NEUTRAL"	"DoubleWoven"	"BLUE"	"Other"
##	L-5	"DoubleWoven"	"Other"	"HandTufted"	"Round"	"Navy"
##	M-2	"PINK"	"Square"	"Rectangle"	"Purple"	"Knotted"
##	RC	"DoubleWoven"	"BLUE"	"Navy"	"NAVY"	"Square"
##	P-4	"DoubleWoven"	"HandTufted"	"Round"	"Navy"	"NAVY"
##	T-4	"Durry"	"HandTufted"	"BLUSHPINK"	"Jacquard"	"Gray"
##	PC	"TAN"	"BLUE"	"Navy"	"NAVY"	"Durry"
##	A-11	"Other"	"Round"	"BLUE"	"Navy"	"NAVY"
##	CC	"Round"	"Handloom"	"Other"	"DoubleWoven"	"HandTufted"
##		[,86]	[,87]	[,88]	[,89]	[,90]
##	H-2	"HandTufted"	"Round"	"DoubleWoven"	"Jacquard"	"TAN"
##	P-5	"Handloom"	"BLUE"	"Jacquard"	"BLUSHPINK"	"Knotted"
##	M-1	"Gray"	"Purple"	"BLUE"	"DoubleBack"	"Durry"
##	A-9	"Handloom"	"Other"	"DoubleWoven"	"Durry"	"HandTufted"
##	C-2	"DoubleBack"	"Purple"	"NEUTRAL"	"DoubleWoven"	"BLUE"
##	JL	"NAVY"	"HandTufted"	"BLUE"	"Handloom"	"Knotted"
##	N-1	"TAN"	"BLUE"	"Navy"	"NAVY"	"Other"
##	T-5	"Square"	"Knotted"	"Rectangle"	"DoubleBack"	"NEUTRAL"
##	C-1	"HandTufted"	"NEUTRAL"	"Gray"	"Square"	"BLUSHPINK"
##	T-2	"HandTufted"	"Square"	"Knotted"	"Gray"	"Purple"
##	I-2	"Navy"	"NAVY"	"Durry"	"DoubleWoven"	"Other"
##	PD	"Round"	"Navy"	"NAVY"	"Handloom"	"HandTufted"
##	L-5	"NAVY"	"Handloom"	"BLUE"	"BLUSHPINK"	"Gray"
##	M-2	"DoubleBack"	"TAN"	"BLUE"	"Navy"	"NAVY"
##	RC	"Handloom"	"Other"	"HandTufted"	"Round"	"Durry"
##	P-4	"Handloom"	"BLUE"	"Other"	"Jacquard"	"TAN"
##	T-4	"PINK"	"NEUTRAL"	"Square"	"Rectangle"	"Knotted"
##	PC	"Round"	"Handloom"	"Other"	"DoubleWoven"	"NEUTRAL"
##	A-11	"HandTufted"	"Jacquard"	"DoubleWoven"	"Durry"	"TAN"
##	CC	"NEUTRAL"	"BLUSHPINK"	"Jacquard"	"Gray"	"Rectangle"
##		[,91]	[,92]	[,93]	[,94]	[,95]
##	H-2	"Rectangle"	"Durry"	"Gray"	"BLUSHPINK"	"Square"
##	P-5	"Durry"	"DoubleBack"	"PINK"	"Purple"	"Rectangle"
##	M-1	"HandTufted"	"Navy"	"NAVY"	"Round"	"Other"
##	A-9	"Jacquard"	"BLUSHPINK"	"NEUTRAL"	"Gray"	"PINK"
##	C-2	"Other"	"HandTufted"	"Handloom"	"Navy"	"NAVY"
##	JL	"Other"	"Round"	"Durry"	"BLUSHPINK"	"Jacquard"
##	N-1	"Round"	"Handloom"	"Durry"	"HandTufted"	"DoubleWoven"
##	T-5	"Purple"	"DoubleWoven"	"TAN"	"BLUE"	"Other"

RECOMMENDATION SYSTEM

## C-1	"PINK"	"Jacquared"	"Purple"	"DoubleBack"	"Rectangle"
## T-2	"Jacquared"	"PINK"	"BLUSHPINK"	"NEUTRAL"	"DoubleBack"
## I-2	"Round"	"HandTufted"	"Handloom"	"Gray"	"Jacquared"
## PD	"Gray"	"Jacquared"	"BLUSHPINK"	"TAN"	"Square"
## L-5	"Jacquared"	"TAN"	"Square"	"PINK"	"Knotted"
## M-2	"Handloom"	"Other"	"DoubleWowen"	"Round"	"Durry"
## RC	"Gray"	"NEUTRAL"	"BLUSHPINK"	"PINK"	"Purple"
## P-4	"BLUSHPINK"	"Gray"	"Durry"	"Square"	"Rectangle"
## T-4	"Purple"	"DoubleBack"	"TAN"	"BLUE"	"Navy"
## PC	"BLUSHPINK"	"PINK"	"HandTufted"	"Gray"	"Jacquared"
## A-11	"Handloom"	"Gray"	"BLUSHPINK"	"Knotted"	"Square"
## CC	"Purple"	"Knotted"	"DoubleBack"	"TAN"	"BLUE"
##	[,96]	[,97]	[,98]	[,99]	[,100]
## H-2	"PINK"	"Knotted"	"NEUTRAL"	"DoubleBack"	"Purple"
## P-5	"NEUTRAL"	"DoubleWowen"	"Square"	"Gray"	"HandTufted"
## M-1	"PINK"	"BLUSHPINK"	"Handloom"	"Jacquared"	"NEUTRAL"
## A-9	"Square"	"Rectangle"	"Purple"	"Knotted"	"DoubleBack"
## C-2	"Round"	"Gray"	"TAN"	"Jacquared"	"BLUSHPINK"
## JL	"Purple"	"PINK"	"DoubleBack"	"NEUTRAL"	"Rectangle"
## N-1	"Jacquared"	"Rectangle"	"NEUTRAL"	"BLUSHPINK"	"Gray"
## T-5	"Round"	"Handloom"	"Navy"	"NAVY"	"HandTufted"
## C-1	"Knotted"	"TAN"	"BLUE"	"Navy"	"NAVY"
## T-2	"Rectangle"	"TAN"	"BLUE"	"Navy"	"NAVY"
## I-2	"BLUSHPINK"	"NEUTRAL"	"Square"	"PINK"	"Purple"
## PD	"PINK"	"Knotted"	"Durry"	"Rectangle"	"DoubleBack"
## L-5	"Durry"	"Rectangle"	"DoubleBack"	"Purple"	"NEUTRAL"
## M-2	"HandTufted"	"BLUSHPINK"	"Gray"	"NEUTRAL"	"Jacquared"
## RC	"Jacquared"	"Knotted"	"DoubleBack"	"Rectangle"	"TAN"
## P-4	"Knotted"	"PINK"	"DoubleBack"	"Purple"	"NEUTRAL"
## T-4	"NAVY"	"Round"	"Other"	"DoubleWowen"	"Handloom"
## PC	"Purple"	"Rectangle"	"Knotted"	"DoubleBack"	"Square"
## A-11	"Purple"	"DoubleBack"	"Rectangle"	"PINK"	"NEUTRAL"
## CC	"Navy"	"NAVY"	"Durry"	"Square"	"PINK"

8. What will be your final recommendation to Champo Carpets?

Answer:

- Some recommendations to Champo carpets would be:
 - [1] Send out more samples and focus on **A-9 and C-1 Customer code** as they are more likely to place order amongst all other customers.
 - [2] Send out more samples to customers from **USA and UK**.
 - [3] Producing and sending out samples of **'Hand Tufted'** and **'Durry'** as these are the highly in demand and have more chances of conversion.
 - [4] Sending out samples with Quality **'TUFTED 60C+VISC 2/16 5PLY'** and Design Name **'ZILLAH'** can be highly effective as it has the highest number of orders.
 - [5] Manufacturing more products in the color **'100 Multi'**.
 - [6] Sending out and manufacturing carpets in **'Rectangular'** shape.