

# Analysis of Sales Report of a Clothes Manufacturing Outlet

Andrei Enescu

10/12/2020

## Background and Objective

A high-end fashion retail store is looking to expand its products. It wants to understand the market and find the current trends in the industry. It has a database of all products with attributes, such as, style, material, season, and the sales of the products over a period of two months.

## Domain:

Retail

## Dataset Description:

There are two files provided: Attribute DataSet.xlsx and Dress Sales.xlsx

## Analysis Tasks:

There are 5 goals / questions for this project

1. To automate the process of recommendations, the store needs to analyze the given attributes of the product, like the style, season, etc., and come up with a model to predict the recommendation of products (in binary output – 0 or 1) accordingly.
2. In order to stock the inventory, the store wants to analyze the sales data and predict the trend of total sales for each dress for an extended period of three more alternative days.
3. To decide the pricing for various upcoming clothes, they wish to find how the style, season, and material affect the sales of a dress and if the style of the dress is more influential than its price.
4. Also, to increase sales, the management wants to analyze the attributes of dresses and find which are the leading factors affecting the sale of a dress.
5. To regularize the rating procedure and find its efficiency, the store wants to find if the rating of the dress affects the total sales

---

These are the packages / libraries used

```
library(plyr)
library(dplyr)
library(rio)
library(raster)
library(data.table)
library(lattice)
library(ggplot2)
library(caret)
library(caTools)
library(forecast)
library(corrplot)
```

First, I imported the two datasets:

1. Attribute DataSet.xlsx contains all the attributes of the dresses and also a recommendation column (binary value suggesting a recommendation (1) or not (0))
2. Dress Sales.xlsx contains the sales for each dress on a particular date. Date ranges from 29/8/2013 to 12/10/2013, and the sales are registered for alternative days.

```
Attrset <- import("Attribute DataSet.xlsx")
Sales.dates <- import("Dress Sales.xlsx")

View(Attrset)
str(Attrset)
```

```
## 'data.frame': 500 obs. of 14 variables:
## $ Dress_ID : num 1.01e+09 1.21e+09 1.19e+09 9.66e+08 8.76e+08 ...
## $ Style : chr "Sexy" "Casual" "vintage" "Brief" ...
## $ Price : chr "Low" "Low" "High" "Average" ...
## $ Rating : num 4.6 0 0 4.6 4.5 0 0 0 0 0 ...
## $ Size : chr "M" "L" "L" "L" ...
## $ Season : chr "Summer" "Summer" "Autumn" "Spring" ...
## $ NeckLine : chr "o-neck" "o-neck" "o-neck" "o-neck" ...
## $ SleeveLength : chr "sleeveless" "Petal" "full" "full" ...
## $ waiseline : chr "empire" "natural" "natural" "natural" ...
## $ Material : chr "null" "microfiber" "polyester" "silk" ...
## $ FabricType : chr "chiffon" "null" "null" "chiffon" ...
## $ Decoration : chr "ruffles" "ruffles" "null" "embroidary" ...
## $ Pattern Type : chr "animal" "animal" "print" "print" ...
## $ Recommendation: num 1 0 0 1 0 0 0 0 1 1 ...
```

## Data Preparation

### Converting from Character to Factor

```
Attrset <- mutate_if(Attrset, is.character, as.factor)
```

Renaming the factor levels as there were mistypes in the original dataset

```
levels(Attrset$Style)
```

```
## [1] "bohemian" "Brief"      "Casual"    "cute"      "fashion"   "Flare"
## [7] "Novelty"  "OL"        "party"     "sexy"      "Sexy"      "vintage"
## [13] "work"
```

```
Attrset$Style <- revalue(Attrset$Style, c("sexy" = "Sexy"))
```

```
levels(Attrset$Price)
```

```
## [1] "Average"    "high"      "High"      "low"       "Low"       "Medium"
## [7] "very-high"
```

```
Attrset$Price <- revalue(Attrset$Price, c("high" = "High"))
```

```
Attrset$Price <- revalue(Attrset$Price, c("low" = "Low"))
```

```
levels(Attrset$Size)
```

```
## [1] "free"  "L"     "M"     "s"     "S"     "small" "XL"
```

```
Attrset$Size <- revalue(Attrset$Size, c("s" = "S"))
```

```
Attrset$Size <- revalue(Attrset$Size, c("small" = "S"))
```

```
levels(Attrset$Season)
```

```
## [1] "Automn" "Autumn" "spring"  "Spring"  "summer"  "Summer"  "winter"  "Winter"
```

```
Attrset$Season <- revalue(Attrset$Season, c("Automn" = "Autumn"))
```

```
Attrset$Season <- revalue(Attrset$Season, c("spring" = "Spring"))
```

```
Attrset$Season <- revalue(Attrset$Season, c("summer" = "Summer"))
```

```
Attrset$Season <- revalue(Attrset$Season, c("winter" = "Winter"))
```

```
levels(Attrset$NeckLine)
```

```
## [1] "backless"      "boat-neck"      "bowneck"         "halter"
## [5] "mandarin-collor" "NULL"           "o-neck"          "open"
## [9] "peterpan-collor" "ruffled"        "Scoop"           "slash-neck"
## [13] "square-collor"  "sweetheart"     "Sweetheart"      "turndowncollor"
## [17] "v-neck"
```

```
Attrset$NeckLine <-
```

```
  revalue(Attrset$NeckLine, c("sweetheart" = "Sweetheart"))
```

```
levels(Attrset$SleeveLength)
```

```
## [1] "butterfly"      "cap-sleeves"    "capsleeves"     "full"
## [5] "half"           "halfsleeve"     "NULL"           "Petal"
## [9] "short"          "sleeveless"     "sleeveless"     "sleeveless"
## [13] "sleeveless"     "threequarter"   "threequarter"    "thressqatar"
## [17] "turndowncollor" "urndowncollor"
```

```

Attrrset$SleeveLength <- revalue(Attrrset$SleeveLength,
                                c("cap-sleeves" = "capsleeves"))
Attrrset$SleeveLength <- revalue(Attrrset$SleeveLength,
                                c("half" = "halfsleeve"))
Attrrset$SleeveLength <- revalue(Attrrset$SleeveLength,
                                c("sleeveless" = "sleeveless"))
Attrrset$SleeveLength <- revalue(Attrrset$SleeveLength,
                                c("sleeveless" = "sleeveless"))
Attrrset$SleeveLength <- revalue(Attrrset$SleeveLength,
                                c("sleeveless" = "sleeveless"))
Attrrset$SleeveLength <- revalue(Attrrset$SleeveLength,
                                c("threequarter" = "threequarter"))
Attrrset$SleeveLength <- revalue(Attrrset$SleeveLength,
                                c("thressqatar" = "threequarter"))
Attrrset$SleeveLength <- revalue(Attrrset$SleeveLength,
                                c("urndowncollor" = "turndowncollor"))

levels(Attrrset$Material)

```

```

## [1] "acrylic"      "cashmere"      "chiffonfabric" "cotton"
## [5] "knitting"     "lace"          "linen"         "lycra"
## [9] "microfiber"   "milksilk"      "mix"           "modal"
## [13] "model"        "null"          "nylon"         "other"
## [17] "polyster"     "rayon"         "shiffon"       "silk"
## [21] "sill"         "spandex"       "viscos"        "wool"

```

```

Attrrset$Material <- revalue(Attrrset$Material, c("model" = "modal"))

levels(Attrrset$FabricType)

```

```

## [1] "batik"      "broadcloth" "chiffon"     "Corduroy"  "dobby"
## [6] "flannael"   "flannel"    "jersey"      "knitted"   "knitting"
## [11] "lace"       "null"       "organza"     "other"     "poplin"
## [16] "satin"      "sattin"     "shiffon"     "terry"     "tulle"
## [21] "wollen"     "woolen"     "worsted"

```

```

Attrrset$FabricType <-
  revalue(Attrrset$FabricType, c("shiffon" = "chiffon"))
Attrrset$FabricType <-
  revalue(Attrrset$FabricType, c("flannael" = "flannel"))
Attrrset$FabricType <-
  revalue(Attrrset$FabricType, c("knitting" = "knitted"))
Attrrset$FabricType <-
  revalue(Attrrset$FabricType, c("sattin" = "satin"))
Attrrset$FabricType <-
  revalue(Attrrset$FabricType, c("wollen" = "woolen"))

levels(Attrrset$Decoration)

```

```

## [1] "applique"  "beading"      "bow"          "button"      "cascading"
## [6] "crystal"   "draped"       "embroidary"   "feathers"     "flowers"
## [11] "hollowout" "lace"         "none"         "null"        "pearls"

```

```
## [16] "plain"      "pleat"      "pockets"    "rivet"      "ruched"
## [21] "ruffles"    "sashes"     "sequined"   "tassel"     "Tiered"
```

```
Attrrset$Decoration <-
  revalue(Attrrset$Decoration, c("none" = "null"))
```

```
levels(Attrrset$`Pattern Type`)
```

```
## [1] "animal"      "character"  "dot"        "floral"     "geometric"  "leopard"
## [7] "leopard"     "none"       "null"       "patchwork"  "plaid"      "print"
## [13] "solid"       "splice"     "striped"
```

```
Attrrset$`Pattern Type` <- revalue(Attrrset$`Pattern Type`,
                                   c("leopard" = "leopard"))
Attrrset$`Pattern Type` <- revalue(Attrrset$`Pattern Type`,
                                   c("none" = "null"))
```

Second dataset - data preparation

```
View(Sales.dates)
```

```
setnames(
  Sales.dates,
  old = c(
    "41314",
    "41373",
    "41434",
    "41495",
    "41556",
    "41617",
    "41315",
    "41374",
    "41435",
    "40400",
    "41557",
    "41618"
  ),
  new = c(
    "2/9/2013",
    "4/9/2013",
    "6/9/2013",
    "8/9/2013",
    "10/9/2013",
    "12/9/2013",
    "2/10/2013",
    "4/10/2013",
    "6/10/2013",
    "8/10/2013",
    "10/10/2013",
    "12/10/2013"
  )
)
```

Replacing the NA values with the mean of the row and also replacing some character values with NA. After Rowmean was calculated I formatted the data frame to have the values without decimals

```
sapply(Sales.dates, class)
```

```
## Dress_ID 29/8/2013 31/8/2013 2/9/2013 4/9/2013 6/9/2013
## "numeric" "numeric" "numeric" "numeric" "numeric" "numeric"
## 8/9/2013 10/9/2013 12/9/2013 14/9/2013 16/9/2013 18/9/2013
## "numeric" "numeric" "character" "character" "character" "character"
## 20/9/2013 22/9/2013 24/9/2013 26/9/2013 28/9/2013 30/9/2013
## "character" "character" "numeric" "numeric" "numeric" "numeric"
## 2/10/2013 4/10/2013 6/10/2013 8/10/2013 10/10/2013 12/10/2013
## "numeric" "numeric" "numeric" "numeric" "numeric" "numeric"
```

```
Sales.dates[Sales.dates == "removed"] <- NA
Sales.dates[Sales.dates == "Removed"] <- NA
Sales.dates[Sales.dates == "Orders"] <- NA

Sales.dates[9:14] <-
  mutate_if(Sales.dates[9:14], is.character, as.numeric)

Sales.dates <-
  (is.na(Sales.dates)) * rowMeans(Sales.dates[2:24], na.rm = T)[row(Sales.dates)] +
  replace(Sales.dates, is.na(Sales.dates), 0)

Sales.dates <-
  Sales.dates %>% mutate_at(vars(-Dress_ID), funs(round(., 0)))
```

Calculating the Total sales from the Sales.dates data frame (both row and column wise); this will be used later in the analysis

```
Totalsales <- rowSums(Sales.dates[2:24])
TotalSalescol <- colSums(Sales.dates[2:24])
Sales.dates <- data.frame(Sales.dates, Totalsales)
```

Merging the two data frames; this will be used later in the analysis

```
TotalIDF <- merge(Attrset, Sales.dates)
```

## Question 1

To automate the process of recommendations, the store needs to analyze the given attributes of the product, like the style, season, etc., and come up with a model to predict the recommendation of products (in binary output – 0 or 1) accordingly.

Bellow I use a logistic regression model. For this the variables are:

1. Dependent variable: Recommendation
2. Remaining variables, except Dress\_ID, which is an identifier

For this model I am creating a new data frame without the Dress\_ID and NA rows

```

Attrib <- Attrset[, -1]
Attrib <- na.omit(Attrib)

fit1 <-
  glm(
    Recommendation ~ Style + Price + Rating + Size + Season +
      NeckLine + SleeveLength + waiseline + Material + FabricType +
      Decoration + 'Pattern Type',
    family = binomial(),
    Attrib
  )
summary(fit1)

##
## Call:
## glm(formula = Recommendation ~ Style + Price + Rating + Size +
##      Season + NeckLine + SleeveLength + waiseline + Material +
##      FabricType + Decoration + 'Pattern Type', family = binomial(),
##      data = Attrib)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.26507  -0.88021  -0.00025   0.87680   2.29953
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -5.216e+01  1.248e+04  -0.004  0.99667
## StyleBrief      -1.607e+00  8.715e-01  -1.844  0.06525 .
## StyleCasual     -4.059e-01  5.592e-01  -0.726  0.46785
## Stylecute       1.289e-01  6.793e-01   0.190  0.84950
## Stylefashion   -2.310e+01  1.130e+04  -0.002  0.99837
## StyleFlare     -1.790e-01  1.695e+00  -0.106  0.91589
## StyleNovelty    6.284e-01  1.032e+00   0.609  0.54242
## StyleOL        1.435e+01  1.130e+04   0.001  0.99899
## Styleparty     -3.687e-01  7.340e-01  -0.502  0.61540
## StyleSexy      -5.087e-01  6.195e-01  -0.821  0.41155
## Stylevintage   -7.849e-01  7.947e-01  -0.988  0.32332
## Stylework      -1.668e+00  9.669e-01  -1.725  0.08445 .
## PriceHigh      -1.105e+00  7.174e-01  -1.541  0.12343
## PriceLow        2.159e-01  2.877e-01   0.750  0.45308
## PriceMedium     1.940e+00  6.269e-01   3.095  0.00197 **
## Pricevery-high  4.133e-01  8.352e-01   0.495  0.62072
## Rating         1.590e-01  6.750e-02   2.356  0.01847 *
## SizeL          -4.403e-01  3.655e-01  -1.205  0.22839
## SizeM          -2.656e-02  3.193e-01  -0.083  0.93371
## SizeS          -6.380e-01  4.990e-01  -1.279  0.20106
## SizeXL         8.005e-03  8.152e-01   0.010  0.99217
## SeasonSpring    1.087e+00  4.014e-01   2.709  0.00675 **
## SeasonSummer   -1.898e-01  3.903e-01  -0.486  0.62681
## SeasonWinter    4.204e-01  3.948e-01   1.065  0.28688
## NeckLineboat-neck 3.588e+01  9.224e+03   0.004  0.99690
## NeckLinebowneck  3.225e+01  9.224e+03   0.003  0.99721
## NeckLinehalter  7.118e+01  1.305e+04   0.005  0.99565

```

## NeckLinemandarin-collor	NA	NA	NA	NA
## NeckLineNULL	3.521e+01	9.224e+03	0.004	0.99695
## NeckLineo-neck	3.403e+01	9.224e+03	0.004	0.99706
## NeckLineopen	3.837e+01	1.305e+04	0.003	0.99765
## NeckLinepeterpan-collor	1.580e+01	9.540e+03	0.002	0.99868
## NeckLineruffled	5.184e+01	1.130e+04	0.005	0.99634
## NeckLineScoop	-1.727e+00	9.224e+03	0.000	0.99985
## NeckLineslash-neck	3.416e+01	9.224e+03	0.004	0.99704
## NeckLinesqare-collor	3.280e+01	9.224e+03	0.004	0.99716
## NeckLineSweetheart	3.471e+01	9.224e+03	0.004	0.99700
## NeckLineturndowncollor	3.310e+01	9.224e+03	0.004	0.99714
## NeckLinev-neck	3.441e+01	9.224e+03	0.004	0.99702
## SleeveLengthcapsleeves	1.875e+01	6.523e+03	0.003	0.99771
## SleeveLengthfull	1.863e+01	6.523e+03	0.003	0.99772
## SleeveLengthhalfssleeve	1.850e+01	6.523e+03	0.003	0.99774
## SleeveLengthNULL	1.699e+01	6.523e+03	0.003	0.99792
## SleeveLengthPetal	1.894e+01	1.130e+04	0.002	0.99866
## SleeveLengthshort	1.758e+01	6.523e+03	0.003	0.99785
## SleeveLengthsleeveless	1.851e+01	6.523e+03	0.003	0.99774
## SleeveLengththreethquarter	1.858e+01	6.523e+03	0.003	0.99773
## SleeveLengthturndowncollor	1.790e+01	6.523e+03	0.003	0.99781
## waiselineempire	-5.595e-01	1.319e+00	-0.424	0.67140
## waiselinenatural	-9.873e-01	1.305e+00	-0.757	0.44923
## waiselinenull	-1.295e+00	1.327e+00	-0.976	0.32928
## waiselineprincess	1.661e+01	6.523e+03	0.003	0.99797
## Materialcashmere	2.065e+01	3.112e+03	0.007	0.99471
## Materialchiffonfabric	1.821e+01	3.112e+03	0.006	0.99533
## Materialcotton	1.925e+01	3.112e+03	0.006	0.99506
## Materialknitting	7.496e-01	7.227e+03	0.000	0.99992
## Materiallace	-2.399e+00	1.172e+04	0.000	0.99984
## Materiallinen	2.489e-01	4.342e+03	0.000	0.99995
## Materiallycra	1.830e+01	3.112e+03	0.006	0.99531
## Materialmicrofiber	1.425e+00	7.227e+03	0.000	0.99984
## Materialmilksilk	1.769e+01	3.112e+03	0.006	0.99547
## Materialmix	1.870e+01	3.112e+03	0.006	0.99521
## Materialmodal	1.854e+01	3.112e+03	0.006	0.99525
## Materialnull	1.919e+01	3.112e+03	0.006	0.99508
## Materialnylon	1.986e+01	3.112e+03	0.006	0.99491
## Materialother	1.420e+00	7.227e+03	0.000	0.99984
## Materialpolyster	1.834e+01	3.112e+03	0.006	0.99530
## Materialrayon	2.042e+01	3.112e+03	0.007	0.99477
## Materialshiffon	1.822e+01	3.112e+03	0.006	0.99533
## Materialsilk	1.866e+01	3.112e+03	0.006	0.99522
## Materialsill	2.301e+00	7.227e+03	0.000	0.99975
## Materialsandex	1.720e+01	3.112e+03	0.006	0.99559
## Materialviscos	2.006e+01	3.112e+03	0.006	0.99486
## Materialwool	5.108e-01	7.227e+03	0.000	0.99994
## FabricTypebroadcloth	-1.974e+01	4.300e+03	-0.005	0.99634
## FabricTypechiffon	-1.944e+01	4.300e+03	-0.005	0.99639
## FabricTypeCorduroy	-3.764e+01	6.274e+03	-0.006	0.99521
## FabricTypedobby	-2.051e+01	4.300e+03	-0.005	0.99620
## FabricTypeflannel	-3.835e+01	6.302e+03	-0.006	0.99514
## FabricTypejersey	-1.842e+01	4.300e+03	-0.004	0.99658
## FabricTypeknitted	-1.954e+01	4.300e+03	-0.005	0.99637



```

## FabricTypelace          3.494e+01  1.209e+04  0.003  0.99769
## FabricTypenull         -1.954e+01  4.300e+03 -0.005  0.99637
## FabricTypeorganza       1.440e+01  1.018e+04  0.001  0.99887
## FabricTypeoother       -2.241e+01  1.018e+04 -0.002  0.99824
## FabricTypepoplin       -2.449e+00  7.813e+03  0.000  0.99975
## FabricTypesatin        -1.830e+01  4.300e+03 -0.004  0.99660
## FabricTypeterry        -2.408e+00  7.813e+03  0.000  0.99975
## FabricTypetulle        -1.941e+01  4.300e+03 -0.005  0.99640
## FabricTypewoolen       -1.690e+01  4.300e+03 -0.004  0.99686
## FabricTypeworsted      -1.945e+01  4.300e+03 -0.005  0.99639
## Decorationbeading       2.213e-01  9.245e-01  0.239  0.81084
## Decorationbow          -4.198e-01  1.110e+00 -0.378  0.70519
## Decorationbutton       -5.367e-01  1.340e+00 -0.400  0.68884
## Decorationcascading    -1.588e+01  6.523e+03 -0.002  0.99806
## Decorationcrystal      1.600e+01  6.523e+03  0.002  0.99804
## Decorationdraped       -1.912e+01  6.523e+03 -0.003  0.99766
## Decorationembroidary    1.689e+00  1.271e+00  1.328  0.18410
## Decorationfeathers      1.836e+01  4.609e+03  0.004  0.99682
## Decorationflowers       7.492e-01  1.507e+00  0.497  0.61899
## Decorationhollowout    -1.022e+00  9.619e-01 -1.063  0.28784
## Decorationlace         5.653e-01  7.324e-01  0.772  0.44026
## Decorationnull         3.733e-01  6.999e-01  0.533  0.59380
## Decorationpearls       -1.777e+01  6.523e+03 -0.003  0.99783
## Decorationplain        8.889e-01  1.918e+00  0.463  0.64302
## Decorationpockets      1.050e+00  1.243e+00  0.845  0.39828
## Decorationrivet        -1.782e+01  3.610e+03 -0.005  0.99606
## Decorationruched        3.459e-01  1.829e+00  0.189  0.85000
## Decorationruffles       5.561e-01  9.480e-01  0.587  0.55747
## Decorationsashes       -2.449e-01  7.734e-01 -0.317  0.75148
## Decorationsequined      8.959e-01  9.609e-01  0.932  0.35112
## Decorationtassel       1.878e+01  6.523e+03  0.003  0.99770
## DecorationTiered       -1.686e+01  6.523e+03 -0.003  0.99794
## 'Pattern Type'character -1.725e+01  6.523e+03 -0.003  0.99789
## 'Pattern Type'dot       1.355e+00  9.964e-01  1.360  0.17392
## 'Pattern Type'floral    -1.931e+01  3.986e+03 -0.005  0.99614
## 'Pattern Type'geometric -1.809e+00  1.546e+00 -1.170  0.24209
## 'Pattern Type'leopard   1.368e-01  1.344e+00  0.102  0.91894
## 'Pattern Type'null      5.639e-01  6.843e-01  0.824  0.40993
## 'Pattern Type'patchwork  2.232e-01  7.307e-01  0.305  0.76004
## 'Pattern Type'plaid     -1.748e+01  3.638e+03 -0.005  0.99617
## 'Pattern Type'print     -3.667e-01  6.858e-01 -0.535  0.59290
## 'Pattern Type'solid     -6.757e-02  6.691e-01 -0.101  0.91956
## 'Pattern Type'splice    -1.704e+01  6.523e+03 -0.003  0.99792
## 'Pattern Type'striped   -5.916e-01  9.402e-01 -0.629  0.52919
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 673.98  on 495  degrees of freedom
## Residual deviance: 484.96  on 372  degrees of freedom
## AIC: 732.96
##
## Number of Fisher Scoring iterations: 17

```

## Observations related to Question 1

In the above model we can see that **Price**, **Season** and **Rating** are the main attributes of the dress that are affecting the recommendation.

Furthermore the Residual deviance is lower than the Null deviance, this means that using this model for prediction will be close to the actual values of recommendation.

I am also testing and validating the model. Splitting the data into 80% Training and 20% Testing

```
set.seed(1)
inTrain <-
  createDataPartition(Attrib$Recommendation, p = 0.8, list = FALSE)
Training <- Attrib[inTrain, ]
Testing <- Attrib[-inTrain, ]

fit2 <-
  glm(
    Recommendation ~ Style + Price + Rating + Size + Season + NeckLine +
      SleeveLength + waiseline + Material + FabricType + Decoration +
      'Pattern Type',
    family = binomial(),
    Training
  )
summary(fit2)
```

```
##
## Call:
## glm(formula = Recommendation ~ Style + Price + Rating + Size +
##      Season + NeckLine + SleeveLength + waiseline + Material +
##      FabricType + Decoration + 'Pattern Type', family = binomial(),
##      data = Training)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.01772  -0.73477  -0.00014   0.67936   2.36359
##
## Coefficients: (3 not defined because of singularities)
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    3.805e+01  1.407e+04  0.003  0.99784
## StyleBrief      -1.499e+00  1.044e+00 -1.436  0.15112
## StyleCasual     -2.839e-01  7.392e-01 -0.384  0.70094
## Stylecute       7.475e-01  9.105e-01  0.821  0.41165
## Stylefashion   -2.040e+01  1.019e+04 -0.002  0.99840
## StyleNovelty    1.741e+00  1.361e+00  1.279  0.20091
## StyleOL        -2.119e+01  6.523e+03 -0.003  0.99741
## Styleparty      6.085e-01  1.015e+00  0.600  0.54884
## StyleSexy      -1.554e-01  8.096e-01 -0.192  0.84781
## Stylevintage   -5.352e-01  1.008e+00 -0.531  0.59540
## Stylework      -2.726e+00  1.243e+00 -2.193  0.02831 *
## PriceHigh      -1.323e+00  8.833e-01 -1.498  0.13410
## PriceLow        1.706e-01  3.552e-01  0.480  0.63094
## PriceMedium     2.259e+00  7.602e-01  2.971  0.00296 **
## Pricevery-high  2.500e+00  1.352e+00  1.850  0.06434 .
## Rating          2.816e-01  8.583e-02  3.281  0.00103 **
```

## SizeL	-5.689e-01	4.435e-01	-1.283	0.19956	
## SizeM	-2.295e-01	3.824e-01	-0.600	0.54833	
## SizeS	-1.834e+00	6.937e-01	-2.644	0.00820	**
## SizeXL	-5.743e-01	1.100e+00	-0.522	0.60150	
## SeasonSpring	1.978e+00	5.047e-01	3.920	8.87e-05	***
## SeasonSummer	5.615e-01	4.962e-01	1.132	0.25777	
## SeasonWinter	8.954e-01	4.960e-01	1.805	0.07106	.
## NeckLinebowneck	-4.879e+00	1.748e+00	-2.792	0.00524	**
## NeckLinehalter	3.610e+01	9.224e+03	0.004	0.99688	
## NeckLinemandarin-collor	NA	NA	NA	NA	
## NeckLineNULL	1.710e+01	6.523e+03	0.003	0.99791	
## NeckLineo-neck	-2.336e+00	8.612e-01	-2.713	0.00668	**
## NeckLineopen	-4.801e-01	7.829e+03	0.000	0.99995	
## NeckLinepeterpan-collor	-1.975e+01	3.128e+03	-0.006	0.99496	
## NeckLineScoop	-3.654e+01	9.224e+03	-0.004	0.99684	
## NeckLineslash-neck	-2.914e+00	1.183e+00	-2.463	0.01377	*
## NeckLinesqare-collor	-2.214e+00	1.954e+00	-1.133	0.25703	
## NeckLineSweetheart	-2.205e+00	1.503e+00	-1.467	0.14247	
## NeckLineturndowncollor	-3.996e+00	1.394e+00	-2.866	0.00416	**
## NeckLinev-neck	-2.155e+00	9.153e-01	-2.354	0.01856	*
## SleeveLengthcapsleeves	1.488e+01	6.523e+03	0.002	0.99818	
## SleeveLengthfull	1.896e+01	6.523e+03	0.003	0.99768	
## SleeveLengthhalfsleeve	1.898e+01	6.523e+03	0.003	0.99768	
## SleeveLengthNULL	-2.007e+01	1.130e+04	-0.002	0.99858	
## SleeveLengthPetal	1.941e+01	1.130e+04	0.002	0.99863	
## SleeveLengthshort	1.773e+01	6.523e+03	0.003	0.99783	
## SleeveLengthsleeveless	1.868e+01	6.523e+03	0.003	0.99771	
## SleeveLengththreequarter	1.859e+01	6.523e+03	0.003	0.99773	
## SleeveLengthturndowncollor	-1.211e+00	9.224e+03	0.000	0.99990	
## waiselineempire	-5.627e+01	1.130e+04	-0.005	0.99603	
## waiselinenatural	-5.745e+01	1.130e+04	-0.005	0.99594	
## waiselinenull	-5.775e+01	1.130e+04	-0.005	0.99592	
## waiselineprincess	-4.123e+01	1.305e+04	-0.003	0.99748	
## Materialcashmere	2.244e+01	2.989e+03	0.008	0.99401	
## Materialchiffonfabric	1.916e+01	2.989e+03	0.006	0.99489	
## Materialcotton	2.002e+01	2.989e+03	0.007	0.99466	
## Materialknitting	2.843e+00	7.175e+03	0.000	0.99968	
## Materiallace	1.581e+00	1.062e+04	0.000	0.99988	
## Materiallinen	1.100e+00	4.045e+03	0.000	0.99978	
## Materiallycra	1.785e+01	2.989e+03	0.006	0.99524	
## Materialmicrofiber	3.218e+00	7.175e+03	0.000	0.99964	
## Materialmilksilk	1.864e+01	2.989e+03	0.006	0.99502	
## Materialmix	2.055e+01	2.989e+03	0.007	0.99451	
## Materialmodal	1.873e+01	2.989e+03	0.006	0.99500	
## Materialnull	2.011e+01	2.989e+03	0.007	0.99463	
## Materialnylon	2.106e+01	2.989e+03	0.007	0.99438	
## Materialother	-2.024e+01	8.380e+03	-0.002	0.99807	
## Materialpolyster	1.891e+01	2.989e+03	0.006	0.99495	
## Materialrayon	2.184e+01	2.989e+03	0.007	0.99417	
## Materialshiffon	1.123e+00	7.175e+03	0.000	0.99988	
## Materialsilk	2.029e+01	2.989e+03	0.007	0.99458	
## Materialsill	2.208e+00	7.175e+03	0.000	0.99975	
## Materialsandex	1.739e+01	2.989e+03	0.006	0.99536	
## Materialviscos	2.091e+01	2.989e+03	0.007	0.99442	

```

## Materialwool      1.722e+00  7.175e+03  0.000  0.99981
## FabricTypebroadcloth -1.970e+01  4.330e+03 -0.005  0.99637
## FabricTypechiffon -1.960e+01  4.330e+03 -0.005  0.99639
## FabricTypeCorduroy -3.690e+01  7.829e+03 -0.005  0.99624
## FabricTypedobby -2.079e+01  4.330e+03 -0.005  0.99617
## FabricTypeflannel -3.866e+01  6.316e+03 -0.006  0.99512
## FabricTypejersey -1.918e+01  4.330e+03 -0.004  0.99646
## FabricTypeknitted -2.056e+01  4.330e+03 -0.005  0.99621
## FabricTypelace 3.127e+01  1.210e+04  0.003  0.99794
## FabricTypenull -1.982e+01  4.330e+03 -0.005  0.99635
## FabricTypeorganza 1.128e+01  1.019e+04  0.001  0.99912
## FabricTypeoother NA NA NA NA
## FabricTypepoplin NA NA NA NA
## FabricTypesatin -1.732e+01  4.330e+03 -0.004  0.99681
## FabricTypeterry -3.479e+00  7.829e+03  0.000  0.99965
## FabricTypetulle -3.798e+01  7.829e+03 -0.005  0.99613
## FabricTypewoolen -1.785e+01  4.330e+03 -0.004  0.99671
## FabricTypeworsted -2.015e+01  4.330e+03 -0.005  0.99629
## Decorationbeading -6.126e-01  1.409e+00 -0.435  0.66375
## Decorationbow -9.982e-01  1.700e+00 -0.587  0.55716
## Decorationbutton -4.136e-01  1.702e+00 -0.243  0.80794
## Decorationcascading -1.460e+01  6.523e+03 -0.002  0.99821
## Decorationcrystal 1.440e+01  6.523e+03  0.002  0.99824
## Decorationdraped -1.967e+01  6.523e+03 -0.003  0.99759
## Decorationembroidary 1.032e+00  1.752e+00  0.589  0.55593
## Decorationfeathers 1.905e+01  3.976e+03  0.005  0.99618
## Decorationflowers 2.928e-02  1.903e+00  0.015  0.98772
## Decorationhollowout -5.412e-01  1.456e+00 -0.372  0.71014
## Decorationlace 1.073e+00  1.163e+00  0.923  0.35623
## Decorationnull 9.547e-01  1.132e+00  0.843  0.39913
## Decorationpearls -1.643e+01  6.523e+03 -0.003  0.99799
## Decorationplain 2.337e+00  2.167e+00  1.078  0.28085
## Decorationpockets 2.810e+00  1.714e+00  1.639  0.10117
## Decorationrivet -1.714e+01  3.741e+03 -0.005  0.99634
## Decorationruched -1.619e+01  6.523e+03 -0.002  0.99802
## Decorationruffles 4.798e-01  1.449e+00  0.331  0.74055
## Decorationsashes -2.093e-01  1.181e+00 -0.177  0.85936
## Decorationsequined 1.042e+00  1.395e+00  0.747  0.45494
## DecorationTiered -1.734e+01  6.523e+03 -0.003  0.99788
## 'Pattern Type'character -1.692e+01  6.523e+03 -0.003  0.99793
## 'Pattern Type'dot 2.086e+00  1.242e+00  1.679  0.09311
## 'Pattern Type'floral -2.012e+01  3.660e+03 -0.005  0.99561
## 'Pattern Type'geometric -1.292e+00  1.793e+00 -0.721  0.47113
## 'Pattern Type'leopard -2.183e-01  1.692e+00 -0.129  0.89736
## 'Pattern Type'null 5.542e-01  8.491e-01  0.653  0.51400
## 'Pattern Type'patchwork 8.089e-01  9.221e-01  0.877  0.38032
## 'Pattern Type'plaid -1.652e+01  3.686e+03 -0.004  0.99642
## 'Pattern Type'print -2.400e-01  8.429e-01 -0.285  0.77587
## 'Pattern Type'solid 1.242e-01  8.213e-01  0.151  0.87978
## 'Pattern Type'striped -4.040e-02  1.131e+00 -0.036  0.97150
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)

```

```
##
## Null deviance: 537.59 on 396 degrees of freedom
## Residual deviance: 337.88 on 280 degrees of freedom
## AIC: 571.88
##
## Number of Fisher Scoring iterations: 17
```

Creating a new column for prediction of recommendation

```
Testing$Predict.recommend <- 0
```

There were some errors after the split, because some levels of the attributes had values only in the Testing set. The values are small and the impact on predication is insignificant so I decided to have the same values for the bellow levels in both, Training and in Testing data sets

```
fit2$xlevels[["Style"]] <-
  union(fit2$xlevels[["Style"]], levels(Testing[["Style"]]))
fit2$xlevels[["NeckLine"]] <-
  union(fit2$xlevels[["NeckLine"]], levels(Testing[["NeckLine"]]))
fit2$xlevels[["Decoration"]] <-
  union(fit2$xlevels[["Decoration"]], levels(Testing[["Decoration"]]))
fit2$xlevels[["Pattern Type"]] <-
  union(fit2$xlevels[["Pattern Type"]], levels(Testing[["Pattern Type"]]))

Pred <- predict(fit2, Testing, type = "response")
```

```
## Warning in predict.lm(object, newdata, se.fit, scale = 1, type = if (type == :
## prediction from a rank-deficient fit may be misleading
```

Setting threshold as 0.7 for the probability

```
Testing$Predict.recommend <- ifelse(Pred >= 0.7, 1, 0)
Testing$Predict.recommend
```

```
## [1] 1 0 1 1 1 0 1 1 0 1 0 1 0 1 1 1 1 0 1 0 1 1 1 0 0 0 0 1 1 0 1 0 0 0 1 1 1
## [39] 0 1 0 0 1 1 0 0 1 1 1 1 0 1 0 0 1 0 1 0 0 1 0 1 1 0 0 1 1 1 1 1 0 1 0 1 0 1
## [77] 0 1 1 0 0 1 1 1 0 0 0 1 1 1 1 1 1 1 1 1 0 0 0
```

Testing by confusion matrix

```
table(Testing$Recommendation,
      Testing$Predict.recommend,
      dnn = list("Actual", "Predicted"))
```

```
##      Predicted
## Actual 0 1
##      0 25 30
##      1 16 28
```

```
conf.matrix <-
  table(Testing$Recommendation, Testing$Predict.recommend)
```

Accuracy and error of the model:

```
sum(conf.matrix)
```

```
## [1] 99
```

```
sum(diag(conf.matrix)) / sum(conf.matrix)
```

```
## [1] 0.5353535
```

```
1 - sum(diag(conf.matrix)) / sum(conf.matrix)
```

```
## [1] 0.4646465
```

## Overall observation

The accuracy of the test model is only 53.54%

---

## Question 2

In order to stock the inventory, the store wants to analyze the sales data and predict the trend of total sales for each dress for an extended period of three more alternative days.

In order to predict the trend of total sales for an extended period of three more days I used the *auto.arima* function.

In this part of the analysis I am using the *TotalSalescol* created in the Data Preparation part.

```
is.vector(TotalSalescol)
```

```
## [1] TRUE
```

```
TotalSalescol
```

```
## 29/8/2013 31/8/2013 2/9/2013 4/9/2013 6/9/2013 8/9/2013 10/9/2013
##      94883      100483      107081      149336      151829      157647      159391
## 12/9/2013 14/9/2013 16/9/2013 18/9/2013 20/9/2013 22/9/2013 24/9/2013
##      166330      169469      172094      174728      179405      184261      185616
## 26/9/2013 28/9/2013 30/9/2013 2/10/2013 4/10/2013 6/10/2013 8/10/2013
##      172853      193734      172008      173339      174193      206334      175684
## 10/10/2013 12/10/2013
##      176625      215533
```

```
Trend.Pred <- ts(TotalSalescol, start = 1, frequency = 7)
summary(Trend.Pred)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  94883  158519  172853  165776  178015  215533
```

```
Trend.Pred <- auto.arima(Trend.Pred)
summary(Trend.Pred)
```

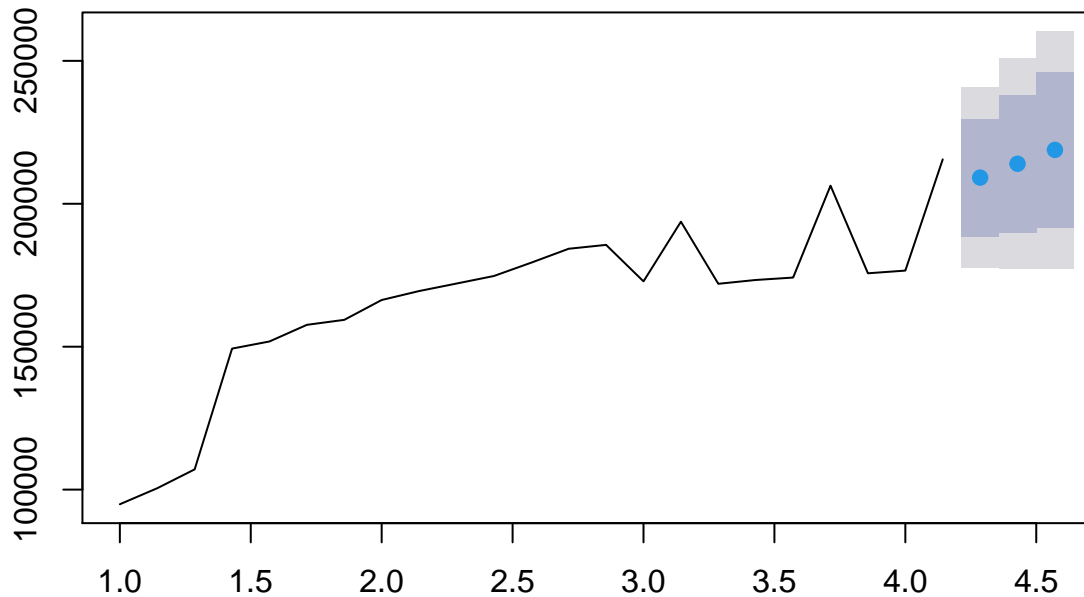
```
## Series: Trend.Pred
## ARIMA(0,1,1) with drift
##
## Coefficients:
##          ma1      drift
##      -0.3944  4846.298
## s.e.    0.1938  2072.032
##
## sigma^2 estimated as 258352544:  log likelihood=-243.32
## AIC=492.64  AICc=493.98  BIC=495.92
##
## Training set error measures:
##              ME      RMSE      MAE      MPE      MAPE      MASE      ACF1
## Training set 191.8268 14988.48 10419.99 0.1309597 5.99801 0.4103329 -0.01880812
```

```
forecast(Trend.Pred, 3)
```

```
##      Point Forecast    Lo 80    Hi 80    Lo 95    Hi 95
## 4.285714      209155.1 188556.3 229754.0 177652.0 240658.3
## 4.428571      214001.4 189920.1 238082.8 177172.3 250830.6
## 4.571429      218847.7 191727.5 245968.0 177370.9 260324.6
```

```
plot(forecast(Trend.Pred, 3))
```

## Forecasts from ARIMA(0,1,1) with drift



## Observation

As we can see from the plot and the ARIMA model output, there is a high fluctuation in the confidence interval

## Question 3

To decide the pricing for various upcoming clothes, they wish to find how the style, season, and material affect the sales of a dress and if the style of the dress is more influential than its price.

In this part of the analysis I am using the *TotalSales* created in the Data Preparation part.

Also I am using the ANOVA model to find the effect of **Style**, **Season** and **Material** on Total Sales.

After ANOVA I am using the linear regression model in order to find the relation of the three attributes on Total Sales and then just for the **Price** and **Style**

```
ANOVA.Style <- aov(TotalDF$Totalsales ~ TotalDF$Style)
ANOVA.Season <- aov(TotalDF$Totalsales ~ TotalDF$Season)
ANOVA.Material <- aov(TotalDF$Totalsales ~ TotalDF$Material)
summary(ANOVA.Style)
```

```
##           Df      Sum Sq   Mean Sq F value Pr(>F)
## TotalDF$Style  11 2.920e+09 265472353    1.48  0.135
## Residuals    538 9.647e+10 179318899
```



```
summary(ANOVA.Season)
```

```
##              Df      Sum Sq   Mean Sq F value Pr(>F)
## TotalDF$Season    3 9.663e+08 322084143   1.782  0.15
## Residuals        544 9.835e+10 180782604
## 2 observations deleted due to missingness
```

```
summary(ANOVA.Material)
```

```
##              Df      Sum Sq   Mean Sq F value Pr(>F)
## TotalDF$Material  22 3.643e+09 165604109   0.91  0.582
## Residuals        526 9.575e+10 182033429
## 1 observation deleted due to missingness
```

```
fit3 <- lm(Totalsales ~ Style + Season + Material, TotalDF)
summary(fit3)
```

```
##
## Call:
## lm(formula = Totalsales ~ Style + Season + Material, data = TotalDF)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -20077   -5684   -2598    1069   135810
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      1587.9     8504.2   0.187  0.8520
## StyleBrief        6654.8     4100.8   1.623  0.1052
## StyleCasual       2165.0     2854.8   0.758  0.4486
## Stylecute         4618.9     3390.0   1.363  0.1736
## Stylefashion     -3169.1    13756.1  -0.230  0.8179
## StyleFlare       -3851.7    10115.5  -0.381  0.7035
## StyleNovelty     -1204.4     4837.4  -0.249  0.8035
## StyleOL         -2547.3    13805.3  -0.185  0.8537
## Styleparty       -1182.1     3355.5  -0.352  0.7248
## StyleSexy         5916.3     3084.5   1.918  0.0557
## Stylevintage      5246.8     3801.2   1.380  0.1681
## Stylework        1917.8     4221.3   0.454  0.6498
## SeasonSpring      1705.6     2030.4   0.840  0.4013
## SeasonSummer     -917.0     1923.1  -0.477  0.6337
## SeasonWinter    -1655.8     1970.0  -0.841  0.4010
## Materialcashmere  -592.7    10397.0  -0.057  0.9546
## Materialchiffonfabric 12204.9     8239.3   1.481  0.1391
## Materialcotton    3044.2     7921.0   0.384  0.7009
## Materialknitting   2569.5    12391.0   0.207  0.8358
## Materiallace      1011.9    15724.1   0.064  0.9487
## Materiallinen     5770.6    11170.8   0.517  0.6057
## Materiallycra     2498.8    11090.6   0.225  0.8218
## Materialmicrofiber 15697.3    11119.9   1.412  0.1587
## Materialmilksilk   5905.0     9921.9   0.595  0.5520
## Materialmix       2448.9     8687.6   0.282  0.7782
```

```
## Materialmodal      -2280.8    12379.5  -0.184    0.8539
## Materialnull       3668.6     7933.5   0.462     0.6440
## Materialnylon      1408.3     9040.0   0.156     0.8763
## Materialother      1907.0    12391.0   0.154     0.8777
## Materialpolyster   3415.7     7981.7   0.428     0.6689
## Materialrayon      3676.2     8939.3   0.411     0.6811
## Materialshiffon    -1303.0    11131.5  -0.117     0.9069
## Materialsilk       184.4      8238.7   0.022     0.9822
## Materialsill       -2908.1    15673.0  -0.186     0.8529
## Materialsbandex    -106.1     9906.2  -0.011     0.9915
## Materialviscos     213.5    12415.2   0.017     0.9863
## Materialwool       -716.0    15617.8  -0.046     0.9635
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 13440 on 510 degrees of freedom
## (3 observations deleted due to missingness)
## Multiple R-squared:  0.07173,    Adjusted R-squared:  0.006207
## F-statistic: 1.095 on 36 and 510 DF,  p-value: 0.3278
```

```
fit4 <- lm(Totalsales ~ Price + Style, TotalDF)
summary(fit4)
```

```
##
## Call:
## lm(formula = Totalsales ~ Price + Style, data = TotalDF)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -12168  -6090  -2943    348  143719
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   3740.51   2705.97   1.382   0.1675
## PriceHigh     -3664.79   3165.86  -1.158   0.2475
## PriceLow       1402.63   1310.78   1.070   0.2851
## PriceMedium   -3226.27   2593.32  -1.244   0.2140
## Pricevery-high -5568.54   3513.34  -1.585   0.1136
## StyleBrief     6690.38   3996.23   1.674   0.0947 .
## StyleCasual    2867.40   2758.94   1.039   0.2991
## Stylecute      7065.36   3256.84   2.169   0.0305 *
## Stylefashion  -3194.51  13654.39  -0.234   0.8151
## StyleFlare    -2072.51   9842.88  -0.211   0.8333
## StyleNovelty  -418.07   4711.49  -0.089   0.9293
## StyleOL       -85.24   13856.34  -0.006   0.9951
## Styleparty     3103.80   3498.00   0.887   0.3753
## StyleSexy      5782.36   2997.04   1.929   0.0542 .
## Stylevintage   6506.45   3742.36   1.739   0.0827 .
## Stylework      3466.16   4131.97   0.839   0.4019
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 13380 on 532 degrees of freedom
## (2 observations deleted due to missingness)
```

```
## Multiple R-squared:  0.04121,    Adjusted R-squared:  0.01418
## F-statistic: 1.524 on 15 and 532 DF,  p-value: 0.09161
```

## Observation

By using the ANOVA model individually on the three attributes we can conclude that Style has the lowest p value (0.135) among the three attributes so it has the most impact in the Total Sales. In the linear regression model *fit3* we can see that the Style Sexy is affecting (in a low manner) the Total Sales. Also for model *fit4*, the Style has a higher impact then the price in the Total Sales of the dress. However both models have a low R and R adjusted values, which means that the models using only these attributes can't be used in predicting the Total Sales.

---

## Question 4

Also, to increase sales, the management wants to analyze the attributes of dresses and find which are the leading factors affecting the sale of a dress.

I am creating a new linear regression models with TotalSales as a dependent variable and the rest (except Dress\_ID) as independent variable.

```
TotalData <- TotalDF[,-c(1, 15:37)]
fit5 <- lm(Totalsales ~ ., TotalData)
summary(fit5)
```

```
##
## Call:
## lm(formula = Totalsales ~ ., data = TotalData)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -34442  -5036  -1087   2345   83797
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    15241.62   25233.85   0.604 0.546159
## StyleBrief       6615.57    3752.51   1.763 0.078631 .
## StyleCasual      2701.59    2603.62   1.038 0.300038
## Stylecute       2648.16    3141.14   0.843 0.399676
## Stylefashion   -9333.60   21741.81  -0.429 0.667930
## StyleFlare     -360.52    9128.12  -0.039 0.968514
## StyleNovelty    1373.93    4496.94   0.306 0.760116
## StyleOL        8699.48   21147.39   0.411 0.681008
## Styleparty     3087.53    3442.69   0.897 0.370319
## StyleSexy       6604.32    2898.17   2.279 0.023180 *
## Stylevintage    6687.10    3533.75   1.892 0.059130 .
## Stylework      1338.08    4005.21   0.334 0.738482
## PriceHigh     -1177.91    3190.11  -0.369 0.712135
## PriceLow        441.33    1328.18   0.332 0.739845
## PriceMedium   -3166.26    2598.40  -1.219 0.223700
```

## Pricevery-high	-2148.82	3885.71	-0.553	0.580552	
## Rating	1289.62	292.37	4.411	1.31e-05	***
## SizeL	5962.77	1677.52	3.555	0.000422	***
## SizeM	1517.27	1487.78	1.020	0.308398	
## SizeS	1858.26	2285.92	0.813	0.416727	
## SizeXL	-159.85	3429.84	-0.047	0.962850	
## SeasonSpring	1320.34	1839.67	0.718	0.473338	
## SeasonSummer	-569.42	1779.04	-0.320	0.749071	
## SeasonWinter	-1651.50	1811.18	-0.912	0.362377	
## NeckLineboat-neck	2661.32	17222.16	0.155	0.877267	
## NeckLinebowneck	4435.72	17537.69	0.253	0.800449	
## NeckLinehalter	5544.20	23342.48	0.238	0.812372	
## NeckLinemandarin-collor	NA	NA	NA	NA	
## NeckLineNULL	-69.36	19581.05	-0.004	0.997175	
## NeckLineo-neck	6620.75	16911.91	0.391	0.695637	
## NeckLineopen	15556.58	24535.34	0.634	0.526394	
## NeckLinepeterpan-collor	5181.20	17485.12	0.296	0.767131	
## NeckLineruffled	143396.32	20677.36	6.935	1.54e-11	***
## NeckLineScoop	1647.37	16881.61	0.098	0.922309	
## NeckLineslash-neck	2836.08	17179.69	0.165	0.868957	
## NeckLinesqare-collor	-779.20	18035.81	-0.043	0.965560	
## NeckLineSweetheart	7493.45	17690.03	0.424	0.672076	
## NeckLineturndowncollor	7542.24	17278.35	0.437	0.662688	
## NeckLinev-neck	6211.88	16987.56	0.366	0.714793	
## SleeveLengthcapsleeves	-31244.75	13164.99	-2.373	0.018077	*
## SleeveLengthfull	-28709.98	12388.40	-2.317	0.020956	*
## SleeveLengthhalfssleeve	-27608.16	12433.01	-2.221	0.026913	*
## SleeveLengthNULL	-21048.85	16017.86	-1.314	0.189533	
## SleeveLengthPetal	11482.57	21132.48	0.543	0.587169	
## SleeveLengthshort	-30219.64	12213.89	-2.474	0.013747	*
## SleeveLengthsleeveless	-31296.91	12297.93	-2.545	0.011287	*
## SleeveLengththreequarter	-26096.38	12485.07	-2.090	0.037199	*
## SleeveLengthturndowncollor	-33833.61	14842.29	-2.280	0.023134	*
## waiselineempire	7679.85	6738.10	1.140	0.255033	
## waiselinenatural	5440.48	6661.84	0.817	0.414583	
## waiselinenull	7368.57	6758.89	1.090	0.276247	
## waiselineprincess	2633.90	15022.92	0.175	0.860908	
## Materialcashmere	407.27	9449.60	0.043	0.965643	
## Materialchiffonfabric	5074.80	7391.75	0.687	0.492745	
## Materialcotton	1729.72	7040.30	0.246	0.806043	
## Materialknitting	52.97	11028.09	0.005	0.996170	
## Materiallace	-10351.76	22224.74	-0.466	0.641616	
## Materiallinen	7127.23	9993.01	0.713	0.476104	
## Materiallycra	4661.37	10099.16	0.462	0.644635	
## Materialmicrofiber	-938.28	13845.49	-0.068	0.946003	
## Materialmilksilk	4367.78	8848.43	0.494	0.621831	
## Materialmix	-184.73	7818.07	-0.024	0.981160	
## Materialmodal	-3340.83	11951.88	-0.280	0.779981	
## Materialnull	3235.72	6994.19	0.463	0.643869	
## Materialnylon	2317.44	8142.30	0.285	0.776078	
## Materialother	4076.03	14075.21	0.290	0.772273	
## Materialpolyster	1327.43	7081.39	0.187	0.851395	
## Materialrayon	3962.12	7936.10	0.499	0.617862	
## Materialshiffon	-4951.20	10090.73	-0.491	0.623917	

## Materialsilk	-1554.79	7328.78	-0.212	0.832093
## Materialsill	-3541.12	13759.49	-0.257	0.797028
## Materialsandex	-3723.38	9090.70	-0.410	0.682321
## Materialviscos	222.06	11291.14	0.020	0.984319
## Materialwool	-1939.29	13648.74	-0.142	0.887080
## FabricTypebroadcloth	3405.25	8660.17	0.393	0.694365
## FabricTypechiffon	65.26	8529.52	0.008	0.993899
## FabricTypeCorduroy	1987.75	11886.05	0.167	0.867266
## FabricTypedobby	-16017.24	11178.14	-1.433	0.152627
## FabricTypeflannel	8468.71	11867.68	0.714	0.475873
## FabricTypejersey	5790.97	9137.64	0.634	0.526589
## FabricTypeknitted	2870.57	12504.73	0.230	0.818546
## FabricTypelace	7877.02	22223.99	0.354	0.723188
## FabricTypenull	-1062.72	8454.90	-0.126	0.900036
## FabricTypeorganza	-4572.08	19284.73	-0.237	0.812708
## FabricTypeoother	3682.75	19083.74	0.193	0.847069
## FabricTypepoplin	-10592.80	14990.09	-0.707	0.480173
## FabricTypesatin	41.02	10043.63	0.004	0.996743
## FabricTypeterry	-6428.51	15238.05	-0.422	0.673334
## FabricTypetulle	5153.42	12761.73	0.404	0.686551
## FabricTypewoolen	-1017.27	10895.38	-0.093	0.925657
## FabricTypeworsted	-799.77	8817.09	-0.091	0.927769
## Decorationbeading	-1933.15	4212.32	-0.459	0.646523
## Decorationbow	3651.07	4597.87	0.794	0.427598
## Decorationbutton	542.68	5433.93	0.100	0.920496
## Decorationcascading	2940.40	11954.91	0.246	0.805835
## Decorationcrystal	4926.03	12445.38	0.396	0.692444
## Decorationdraped	3915.34	12054.11	0.325	0.745484
## Decorationembroidary	5370.20	6616.93	0.812	0.417488
## Decorationfeathers	-1958.20	9085.03	-0.216	0.829449
## Decorationflowers	-788.57	7115.96	-0.111	0.911814
## Decorationhollowout	-277.64	4004.39	-0.069	0.944756
## Decorationlace	2408.79	3239.42	0.744	0.457541
## Decorationnull	686.05	3060.43	0.224	0.822736
## Decorationpearls	5134.49	12742.00	0.403	0.687184
## Decorationplain	-3177.41	9909.30	-0.321	0.748635
## Decorationpockets	-1602.86	6177.41	-0.259	0.795398
## Decorationrivet	1475.73	7578.03	0.195	0.845692
## Decorationruched	-5572.74	8316.23	-0.670	0.503159
## Decorationruffles	6935.31	4220.56	1.643	0.101084
## Decorationsashes	3438.46	3453.17	0.996	0.319948
## Decorationsequined	-2683.87	4389.39	-0.611	0.541234
## Decorationtassel	-6543.68	12305.32	-0.532	0.595161
## DecorationTiered	31109.41	13776.87	2.258	0.024451 *
## 'Pattern Type'character	-1609.37	11801.41	-0.136	0.891593
## 'Pattern Type'dot	-1857.61	4564.51	-0.407	0.684238
## 'Pattern Type'floral	949.63	8719.14	0.109	0.913323
## 'Pattern Type'geometric	14286.73	6377.53	2.240	0.025601 *
## 'Pattern Type'leopard	-2897.95	6077.86	-0.477	0.633749
## 'Pattern Type'null	-5901.52	3103.46	-1.902	0.057906 .
## 'Pattern Type'patchwork	-4335.86	3360.15	-1.290	0.197628
## 'Pattern Type'plaid	-6456.94	7479.72	-0.863	0.388486
## 'Pattern Type'print	-2605.65	3147.40	-0.828	0.408212
## 'Pattern Type'solid	-2149.40	3028.17	-0.710	0.478220

```
## 'Pattern Type'splice      -5723.22    9554.25   -0.599 0.549480
## 'Pattern Type'striped     -3087.51    4117.66   -0.750 0.453782
## Recommendation            907.02     1168.51    0.776 0.438055
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 11310 on 421 degrees of freedom
## (4 observations deleted due to missingness)
## Multiple R-squared:  0.4578, Adjusted R-squared:  0.2981
## F-statistic: 2.867 on 124 and 421 DF, p-value: 1.271e-15
```

## Observation

We can see in the output of the model that we have a lot of variables affecting the Total Sales. We can split the variables in high and moderate impact and also positive and negative impact.

1. High - positive impact variables

- a) Rating
- b) Size L
- c) NeckLine ruffled

2. Moderate - positive impact variables

- a) Style Sexy
- b) Decoration Tiered
- c) Pattern Type geometric

3. Moderate - negative impact

- a) SleeveLength

The multiple and adjusted R-squared values are 46% and 30% respectively. We can conclude that the model is quite fit and the p-value of almost 0 also suggests that there is definitely an impact of these variables on the total sales of the dresses.

---

## Question 5

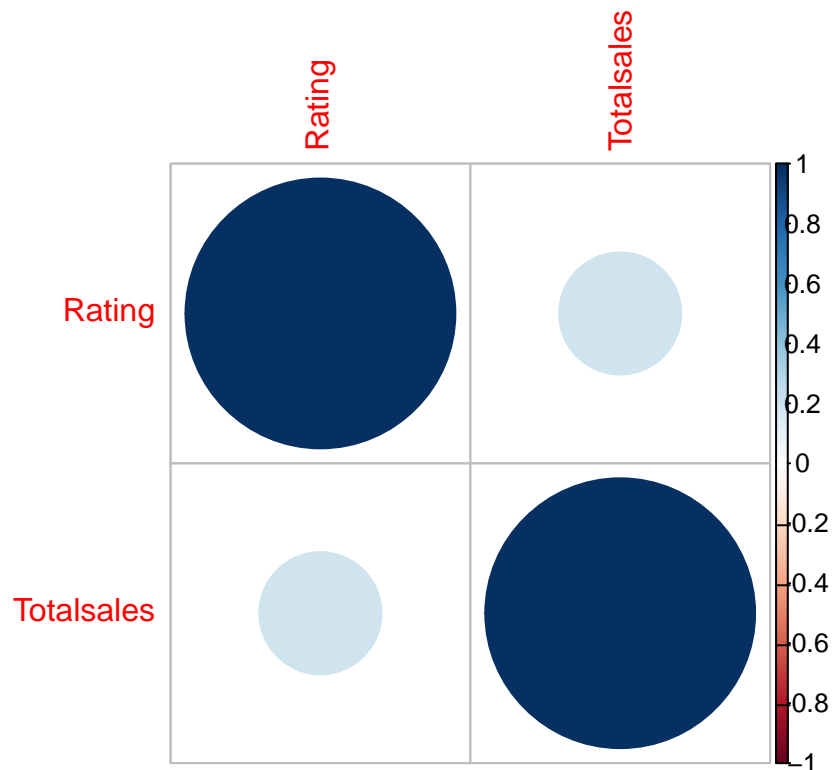
To regularize the rating procedure and find its efficiency, the store wants to find if the rating of the dress affects the total sales

In order to find if there is a relation between *Rating* and *Total Sales*, I use *cor.test* function.

```
corr <- cor.test(TotalDF$Totalsales, TotalDF$Rating)
corr
```

```
##
## Pearson's product-moment correlation
##
## data: TotalDF$Totalsales and TotalDF$Rating
## t = 4.8992, df = 548, p-value = 1.269e-06
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
##  0.1233527 0.2835959
## sample estimates:
##      cor
## 0.2048465
```

```
corr4plot <- cor(TotalDF[, c(4, 38)])
corrplot(corr4plot)
```



## Observation

Based on the output of this function, we have a correlation of 0.2 between the Total Sales and the Rating. This means that there is a weak - positive correlation between the two attributes. In conclusion a higher rating will correlate with higher sales which suggests that the rating approach requires regularization.