

WallMart Coursework in R

Code ▾

Installing Packages and Libraries

Hide

```
install.packages("data.table")
library(data.table)
install.packages("ggplot2")
library(ggplot2)
install.packages("psych")
library(psych)
install.packages("corrplot")
library(corrplot)
install.packages("dplyr")
library(dplyr)
install.packages("plyr")
library(plyr)
install.packages("Amelia")
library(Amelia)
library(tidyr)
library(stringr)
library(dummies)
```

Setting the Working Directory & Loading the datasets as data frames

Hide

```
#setwd("D:/RCW/")
train <- read.csv("D:/RCW/Train.csv")
test = read.csv("D:/RCW/Test.csv")
```

Printing the top 6 rows of the train & test data frames by using head()

Hide

```
head(train)
```

Item_Identifier <fctr>	Item_Weight <dbl>	Item_Fat_Content <fctr>	Item_Visibility <dbl>	Item_Type <fctr>
1 FDA15	9.300	Low Fat	0.01604730	Dairy
2 DRC01	5.920	Regular	0.01927822	Soft Drinks
3 FDN15	17.500	Low Fat	0.01676007	Meat
4 FDX07	19.200	Regular	0.00000000	Fruits and Vegetables
5 NCD19	8.930	Low Fat	0.00000000	Household
6 FDP36	10.395	Regular	0.00000000	Baking Goods
6 rows 1-7 of 12 columns				

Hide

```
head(test)
```

Item_Identifier <fctr>	Item_Weight <dbl>	Item_Fat_Content <fctr>	Item_Visibility <dbl>	Item_Type <fctr>
1 FDW58	20.750	Low Fat	0.007564836	Snack Foods
2 FDW14	8.300	reg	0.038427677	Dairy
3 NCN55	14.600	Low Fat	0.099574908	Others
4 FDQ58	7.315	Low Fat	0.015388393	Snack Foods
5 FDY38	NA	Regular	0.118599314	Dairy
6 FDH56	9.800	Regular	0.063817206	Fruits and Vegetables

6 rows | 1-7 of 11 columns

Checking the Dimensions of the dataset by using dim() The train dataset has 8523(rows) 12(cols) The test dataset has 5681(rows) 11(cols)

Hide

```
dim(train)
```

```
[1] 8523  12
```

Hide

```
dim(test)
```

```
[1] 5681  11
```

Checking the column names to find the missing columns We find the 'Item_Outlet_Sales' is missing from test dataset. This is because we will be predicting the values of 'Item_Outlet_Sales'

Hide

```
names(train)
```

```
[1] "Item_Identifier"      "Item_Weight"          "Item_Fat_Content"
[4] "Item_Visibility"     "Item_Type"            "Item_MRP"
[7] "Outlet_Identifier"    "Outlet_Establishment_Year" "Outlet_Size"
[10] "Outlet_Location_Type" "Outlet_Type"          "Item_Outlet_Sales"
```

Hide

```
names(test)
```

```
[1] "Item_Identifier"      "Item_Weight"      "Item_Fat_Content"
[4] "Item_Visibility"      "Item_Type"        "Item_MRP"
[7] "Outlet_Identifier"    "Outlet_Establishment_Year" "Outlet_Size"
[10] "Outlet_Location_Type" "Outlet_Type"
```

Checking if this data has missing values. We are using table to group the values by False and True. We find 1463 NA values in the train dataset.

[Hide](#)

```
table(is.na(train))
```

```
FALSE  TRUE
100813 1463
```

Checking the variables with the count of NA values. We find only Item_Weight has the 1463 NA values.

[Hide](#)

```
colSums(is.na(train))
```

```

Item_Identifier      Item_Weight      Item_Fat_Content      Item_Vis
ibility              0              1463              0
0
Item_Type            Item_MRP      Outlet_Identifier Outlet_Establishme
nt_Year              0              0              0
0
Outlet_Size      Outlet_Location_Type      Outlet_Type      Item_Outle
t_Sales              0              0              0
0
```

Checking the variables and their types in train dataset

[Hide](#)

```
str(train)
```

```
'data.frame': 8523 obs. of 12 variables:
 $ Item_Identifier      : Factor w/ 1559 levels "DRA12","DRA24",...: 157 9 663 1122 1298 759
697 739 441 991 ...
 $ Item_Weight          : num  9.3 5.92 17.5 19.2 8.93 ...
 $ Item_Fat_Content      : Factor w/ 5 levels "LF","low fat",...: 3 5 3 5 3 5 5 3 5 5 ...
 $ Item_Visibility       : num  0.016 0.0193 0.0168 0 0 ...
 $ Item_Type            : Factor w/ 16 levels "Baking Goods",...: 5 15 11 7 10 1 14 14 6 6
...
 $ Item_MRP             : num  249.8 48.3 141.6 182.1 53.9 ...
 $ Outlet_Identifier     : Factor w/ 10 levels "OUT010","OUT013",...: 10 4 10 1 2 4 2 6 8 3
...
 $ Outlet_Establishment_Year: int  1999 2009 1999 1998 1987 2009 1987 1985 2002 2007 ...
 $ Outlet_Size           : Factor w/ 4 levels "", "High", "Medium",...: 3 3 3 1 2 3 2 3 1 1 ...
 $ Outlet_Location_Type   : Factor w/ 3 levels "Tier 1", "Tier 2",...: 1 3 1 3 3 3 3 3 2 2 ...
 $ Outlet_Type            : Factor w/ 4 levels "Grocery Store",...: 2 3 2 1 2 3 2 4 2 2 ...
 $ Item_Outlet_Sales      : num  3735 443 2097 732 995 ...
```

We will see a summary of the train dataset.

[Hide](#)

```
summary(train)
```

Item_Identifier	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type
FDG33 : 10	Min. : 4.555	LF : 316	Min. :0.00000	Fruits and Vegetables:1232
FDW13 : 10	1st Qu.: 8.774	low fat: 112	1st Qu.:0.02699	Snack Foods :1200
DRE49 : 9	Median :12.600	Low Fat:5089	Median :0.05393	Household : 910
DRN47 : 9	Mean :12.858	reg : 117	Mean :0.06613	Frozen Foods : 856
FDD38 : 9	3rd Qu.:16.850	Regular:2889	3rd Qu.:0.09459	Dairy : 682
FDF52 : 9	Max. :21.350		Max. :0.32839	Canned : 649
(Other):8467	NA's :1463			(Other) :2994

Item_MRP	Outlet_Identifier	Outlet_Establishment_Year	Outlet_Size	Outlet_Location_Type
Min. : 31.29	OUT027 : 935	Min. :1985	:2410	Tier 1:2388
1st Qu.: 93.83	OUT013 : 932	1st Qu.:1987	High : 932	Tier 2:2785
Median :143.01	OUT035 : 930	Median :1999	Medium:2793	Tier 3:3350
Mean :140.99	OUT046 : 930	Mean :1998	Small :2388	
3rd Qu.:185.64	OUT049 : 930	3rd Qu.:2004		
Max. :266.89	OUT045 : 929	Max. :2009		
	(Other):2937			

Outlet_Type	Item_Outlet_Sales
Grocery Store :1083	Min. : 33.29
Supermarket Type1:5577	1st Qu.: 834.25
Supermarket Type2: 928	Median : 1794.33
Supermarket Type3: 935	Mean : 2181.29
	3rd Qu.: 3101.30
	Max. :13086.97

From the above information we can:

Exploring the Numerical Columns:

- 1.Item_Weight - There are 1463 NA Values
- 2.Item Visibility - Contains no NA Values, but contains 0 values.
- 3.Item_MRP - Contains No NA/0 values.Also has an acceptable price range with no outliers.
- 4.Outlet_Establishment_Year - Contains no NA/0 values. Average mean is 1997, implying mostly old stores.
- 5.Item_Outlet_Sales - Contains no NA/0 values.

Exploring the Factor Columns:

- 1.Item_Identifier - Contains 1559 unique values

Hide

```
#install.packages("plyr")
#library(plyr)
#library(dplyr)
train %>%
  summarise(n_distinct(Item_Identifier))
```

n_distinct(Item_Identifier)

<int>

1559

1 row

2.Item_Fat_Content - We find the level values Low Fat/low fat/LF are same but typed incorrectly.

Hide

```
train %>%
  group_by(Item_Fat_Content) %>% summarise(Count = n()) %>% arrange(desc(Count))
```

Item_Fat_Content

<fctr>

Count

<int>

Low Fat

5089

Regular

2889

LF

316

reg

117

low fat

112

5 rows

3.Item_Type - Categories of Items with counts

Hide

```
train%>%
  group_by(Item_Type) %>%
  summarise(Count = n()) %>% arrange(desc(Count))
```

Item_Type

<fctr>

Count

<int>

Fruits and Vegetables

1232

Snack Foods

1200

Household

910

Frozen Foods

856

Dairy

682

Item_Type <fctr>	Count <int>
Canned	649
Baking Goods	648
Health and Hygiene	520
Soft Drinks	445
Meat	425
1-10 of 16 rows	Previous 1 2 Next

4.Outlet_Identifier - There are Item information from 10 different Outlets

[Hide](#)

```
train %>%
  group_by(Outlet_Identifier) %>%
  summarise(Count = n()) %>% arrange(desc(Count))
```

Outlet_Identifier <fctr>	Count <int>
OUT027	935
OUT013	932
OUT035	930
OUT046	930
OUT049	930
OUT045	929
OUT018	928
OUT017	926
OUT010	555
OUT019	528
1-10 of 10 rows	

5.Outlet_Size - Outlet Size data not properly levelled. (2410 counts)

[Hide](#)

```
train%>%
  group_by(Outlet_Size) %>% summarise(Count = n())
```

Outlet_Size <fctr>	Count <int>
-----------------------	----------------

Outlet_Size <fctr>	Count <int>
	2410
High	932
Medium	2793
Small	2388
4 rows	

6.Outlet_Location_Type - Number of Outlet Location type with counts. We find the data is normally distributed.

[Hide](#)

```
train%>%
group_by(Outlet_Location_Type) %>%
summarise(Count = n()) %>% arrange(desc(Count))
```

Outlet_Location_Type <fctr>	Count <int>
Tier 3	3350
Tier 2	2785
Tier 1	2388
3 rows	

7.Outlet_Type - We find the Types of Outlet

[Hide](#)

```
train%>%
group_by(Outlet_Type)%>%
summarise(Count=n())%>% arrange(desc(Count))
```

Outlet_Type <fctr>	Count <int>
Supermarket Type1	5577
Grocery Store	1083
Supermarket Type3	935
Supermarket Type2	928
4 rows	

Data Manipulation

We are creating a new variable in test dataset Item_Outlet_Sales, to match our number of rows with train dataset.

Hide

```
test$Item_Outlet_Sales <- 1
names(test)
```

```
[1] "Item_Identifier"      "Item_Weight"      "Item_Fat_Content"
[4] "Item_Visibility"     "Item_Type"        "Item_MRP"
[7] "Outlet_Identifier"    "Outlet_Establishment_Year" "Outlet_Size"
[10] "Outlet_Location_Type" "Outlet_Type"      "Item_Outlet_Sales"
```

Now, we are combining the train and test data with rbind function

Hide

```
combi <- rbind(train, test)
dim(combi)
```

```
[1] 14204    12
```

We are imputing the NA values in Item_Weight with the median of the values of the column. To calculate the median of the non-missing values if we are passing the argument na.rm=TRUE

Hide

```
combi$Item_Weight[is.na(combi$Item_Weight)] <- median(combi$Item_Weight, na.rm = TRUE)
summary(combi$Item_Weight)
```

```
   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
 4.555   9.300  12.600  12.760  16.000  21.350
```

There are 0 values in Item_visibility, therefore, we also impute the 0s with median of the column values.

Hide

```
combi$Item_Visibility <- ifelse(combi$Item_Visibility == 0, median(combi$Item_Visibility),combi
$Item_Visibility)
summary(combi$Item_Visibility)
```

```
   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.003575 0.033143 0.054023 0.069296 0.094037 0.328391
```

Renaming the blank level in of Outlet_Size to 'Other'

Hide

```
levels(combi$Outlet_Size)[1] <- "Other"
table(combi$Outlet_Size)
```

```
Other    High Medium Small
4016    1553   4655  3980
```

Renaming the levels of Item_Fat_Content to 'Low Fat' & 'Regular'

[Hide](#)

```
#library(plyr)
#combi$Item_Fat_Content <- revalue(combi$Item_Fat_Content,c("LF" = "Low Fat", "reg" = "Regular",
  "low fat" = "Low Fat"))
table(combi$Item_Fat_Content)
```

```
Low Fat Regular
9185      5019
```

Data Visualisation

[Hide](#)

```
combi_encoded=as.data.frame(combi)
str(combi_encoded)
```

```
'data.frame':  14204 obs. of  12 variables:
 $ Item_Identifier      : Factor w/ 1559 levels "DRA12","DRA24",...: 157 9 663 1122 1298 759
697 739 441 991 ...
 $ Item_Weight          : num  9.3 5.92 17.5 19.2 8.93 ...
 $ Item_Fat_Content     : Factor w/ 2 levels "Low Fat","Regular": 1 2 1 2 1 2 2 1 2 2 ...
 $ Item_Visibility      : num  0.016 0.0193 0.0168 0.054 0.054 ...
 $ Item_Type            : Factor w/ 16 levels "Baking Goods",...: 5 15 11 7 10 1 14 14 6 6
...
 $ Item_MRP             : num  249.8 48.3 141.6 182.1 53.9 ...
 $ Outlet_Identifier    : Factor w/ 10 levels "OUT010","OUT013",...: 10 4 10 1 2 4 2 6 8 3
...
 $ Outlet_Establishment_Year: int  1999 2009 1999 1998 1987 2009 1987 1985 2002 2007 ...
 $ Outlet_Size          : Factor w/ 4 levels "Other","High",...: 3 3 3 1 2 3 2 3 1 1 ...
 $ Outlet_Location_Type  : Factor w/ 3 levels "Tier 1","Tier 2",...: 1 3 1 3 3 3 3 3 2 2 ...
 $ Outlet_Type           : Factor w/ 4 levels "Grocery Store",...: 2 3 2 1 2 3 2 4 2 2 ...
 $ Item_Outlet_Sales     : num  3735 443 2097 732 995 ...
```

Dividing data to Train & Test before Label & Hot Encoding

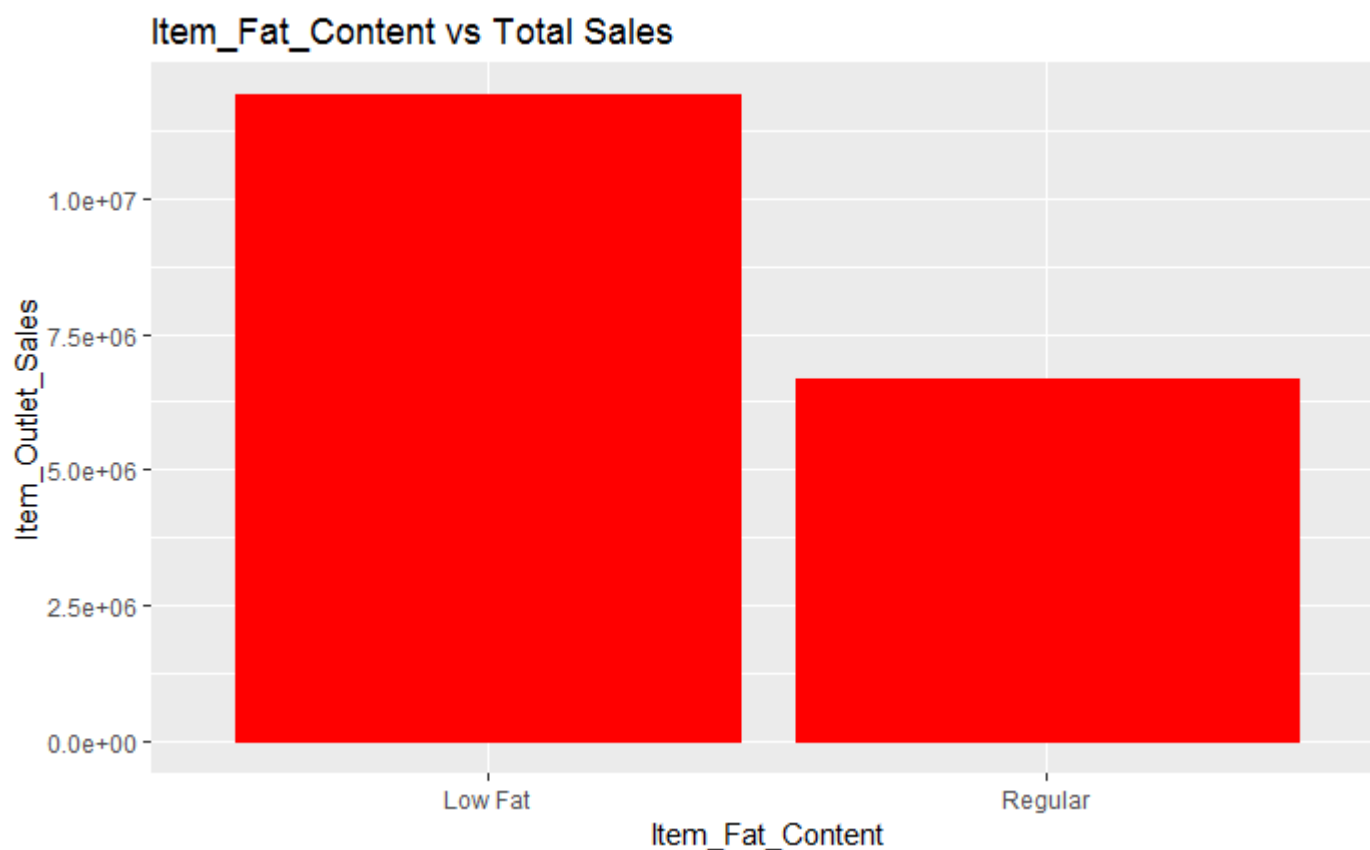
[Hide](#)

```
new_train_combi <- combi %>% filter(Item_Outlet_Sales != 1)
new_test_combi <- combi %>% filter(Item_Outlet_Sales == 1)
str(new_train_combi)
str(new_test_combi)
```

We have tried to visualise Item_Outlet_Sales with different Categorical Values

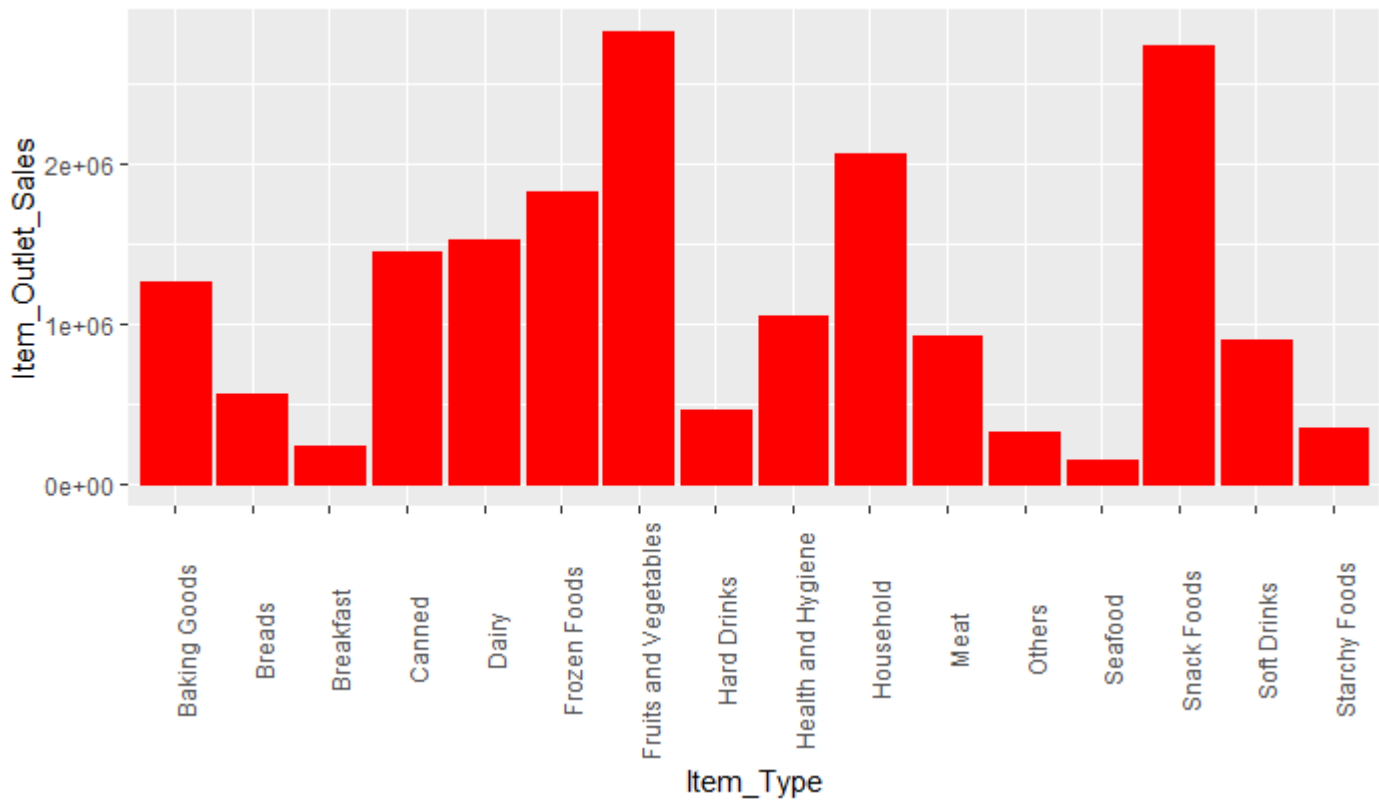
[Hide](#)

```
ggplot(new_train_combi, aes(Item_Fat_Content, Item_Outlet_Sales)) + geom_bar(stat = "identity",  
  color = "red") + ggtitle("Item_Fat_Content vs Total Sales")
```

[Hide](#)

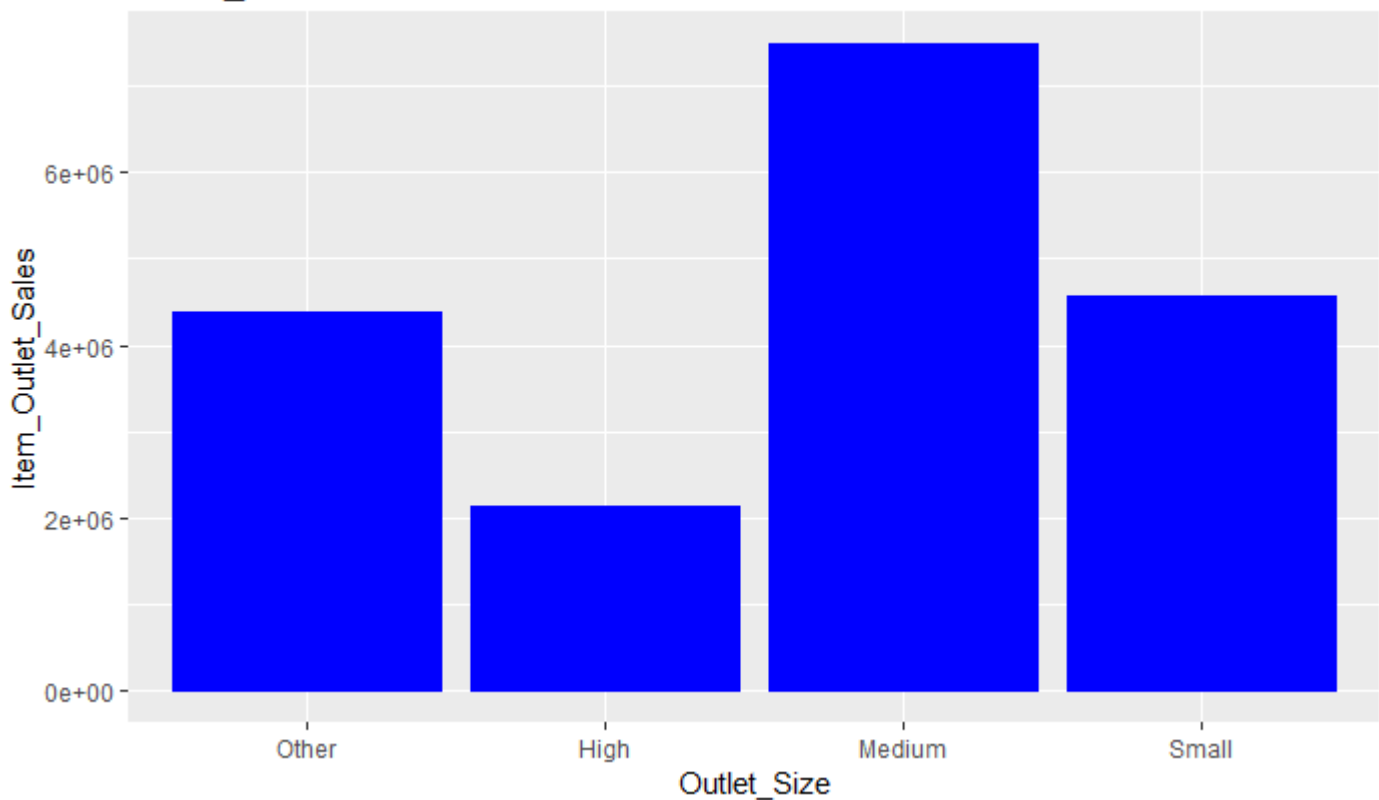
```
ggplot(new_train_combi, aes(Item_Type, Item_Outlet_Sales)) + geom_bar(stat = "identity", color =  
  "red") + theme(axis.text.x = element_text(angle = 90), axis.text.y = element_text(angle = 0)) +  
  ggtitle("Item_Type vs Total Sales")
```

Item_Type vs Total Sales

[Hide](#)

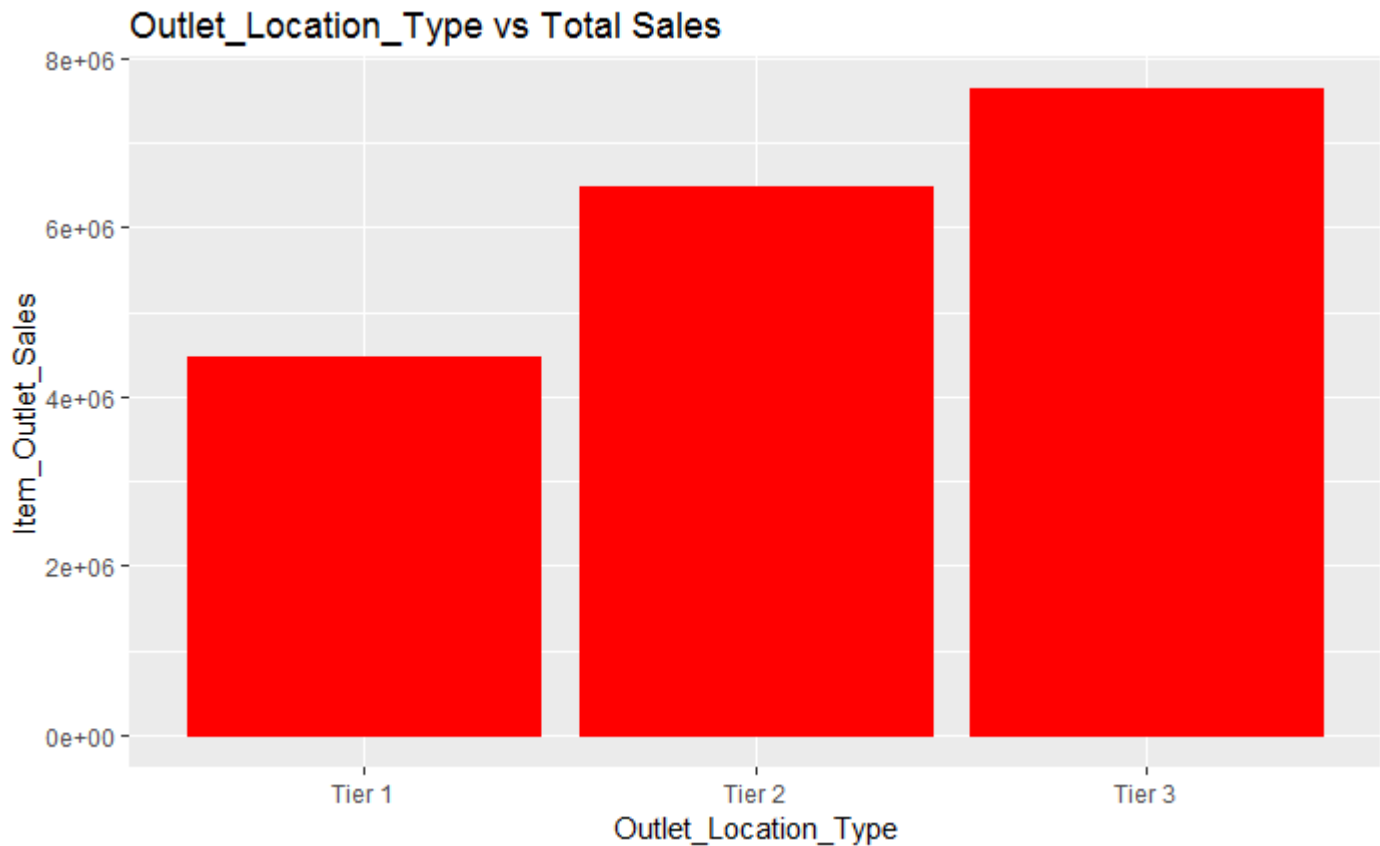
```
ggplot(new_train_combi, aes(Outlet_Size, Item_Outlet_Sales)) + geom_bar(stat = "identity", color = "blue") + ggtitle("Outlet_Size vs Total Sales")
```

Outlet_Size vs Total Sales



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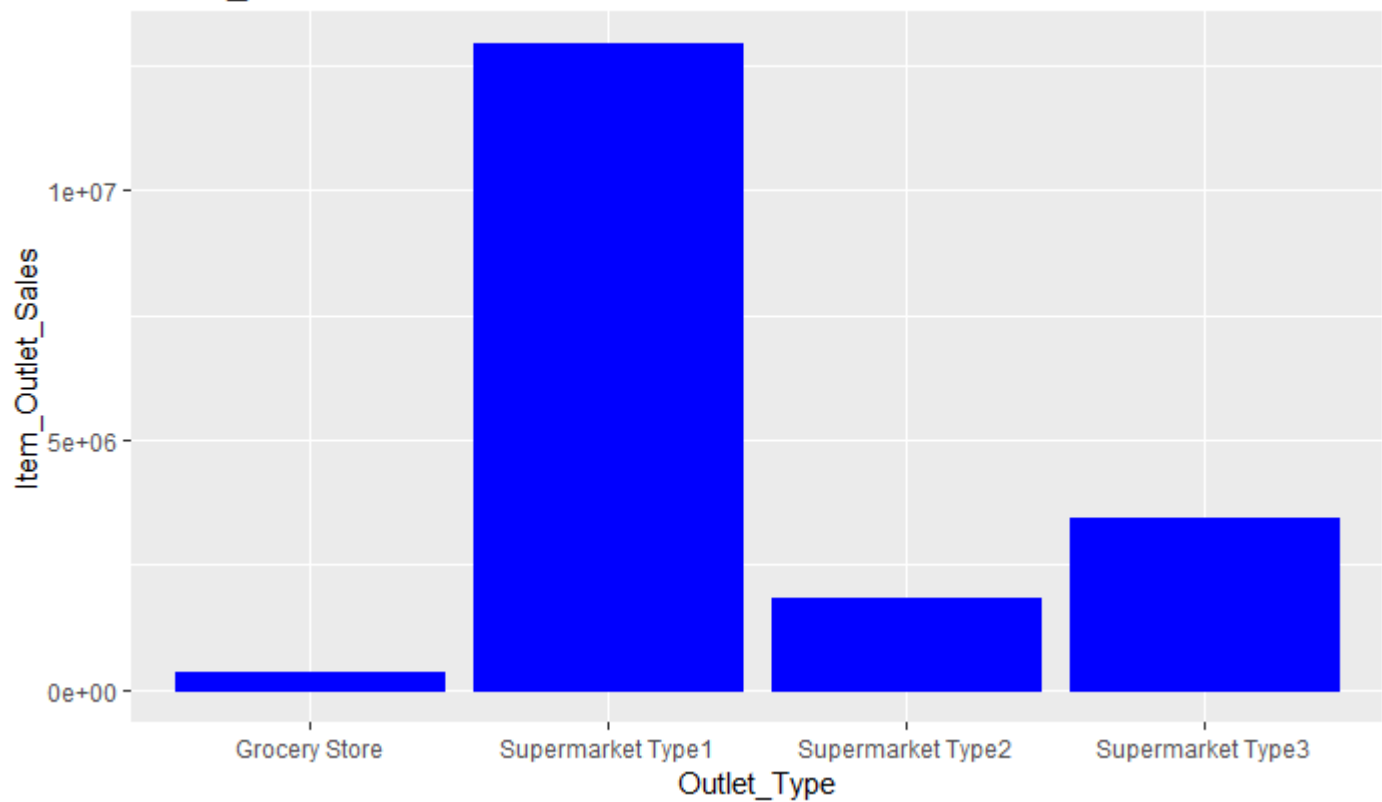
```
ggplot(new_train_combi, aes(Outlet_Location_Type, Item_Outlet_Sales)) + geom_bar(stat = "identity", color = "red") + ggtitle("Outlet_Location_Type vs Total Sales")
```



Hide

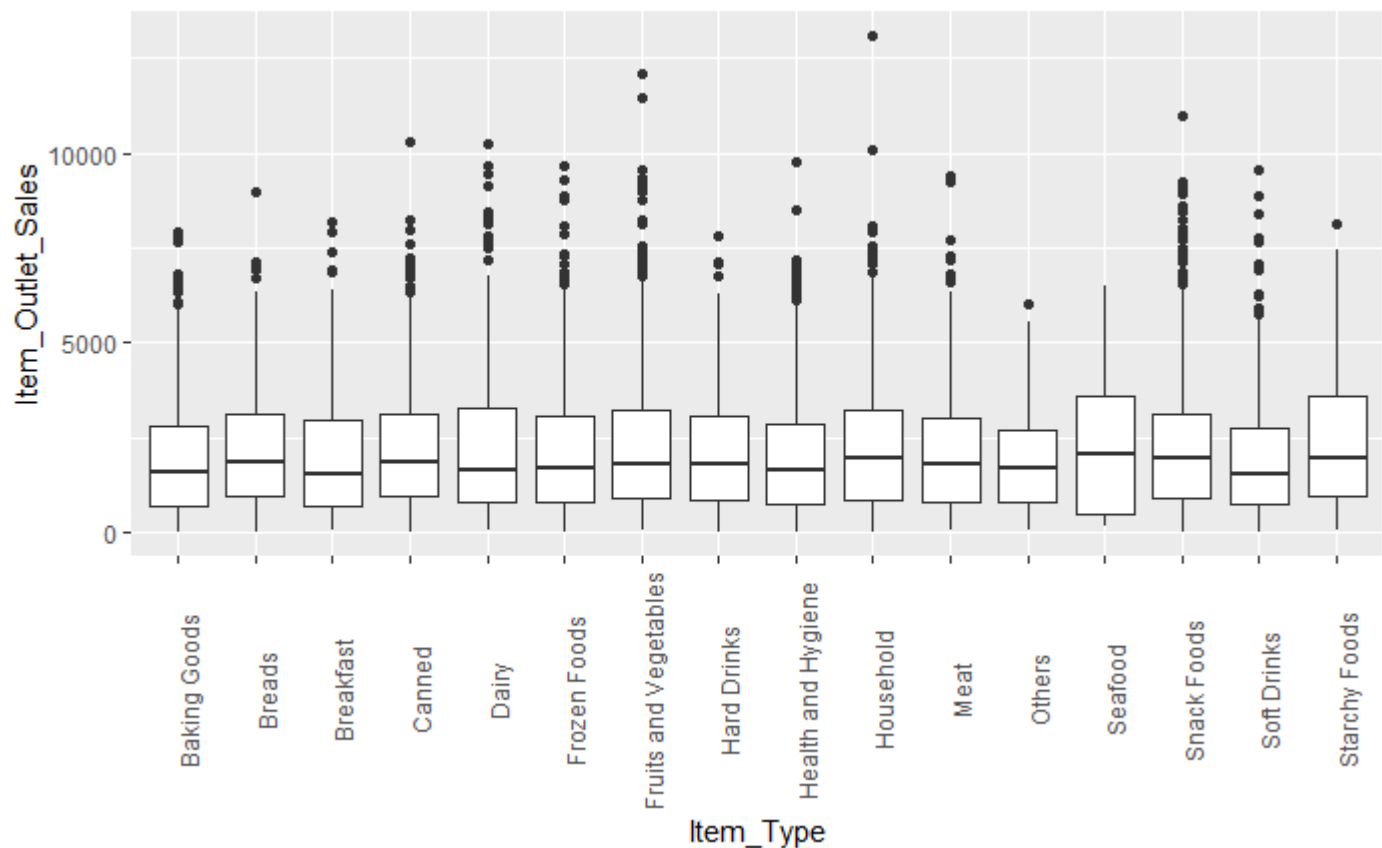
```
ggplot(new_train_combi, aes(Outlet_Type, Item_Outlet_Sales)) + geom_bar(stat = "identity", color = "blue") + ggtitle("Outlet_Size vs Total Sales")
```

Outlet_Size vs Total Sales



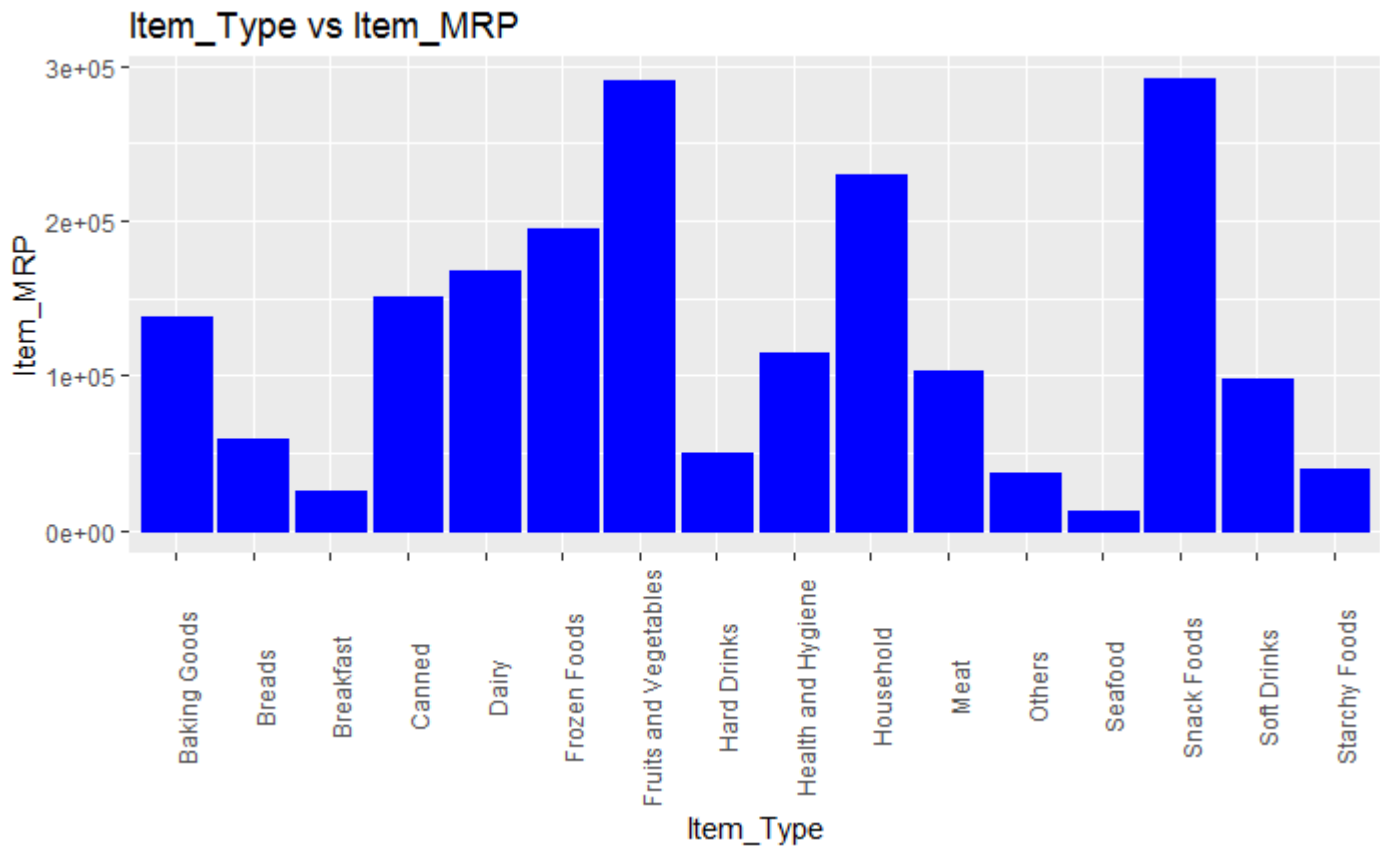
Hide

```
qplot(x=Item_Type,y=Item_Outlet_Sales, data=new_train_combi,geom = "boxplot",) + theme(axis.tex
t.x = element_text(angle = 90), axis.text.y = element_text(angle = 0))
```



Hide

```
ggplot(combi, aes(Item_Type, Item_MRP)) + geom_bar(stat = "identity", color = "blue")+ theme(axis.text.x = element_text(angle = 90), axis.text.y = element_text(angle = 0)) + ggtitle("Item_Type vs Item_MRP")
```

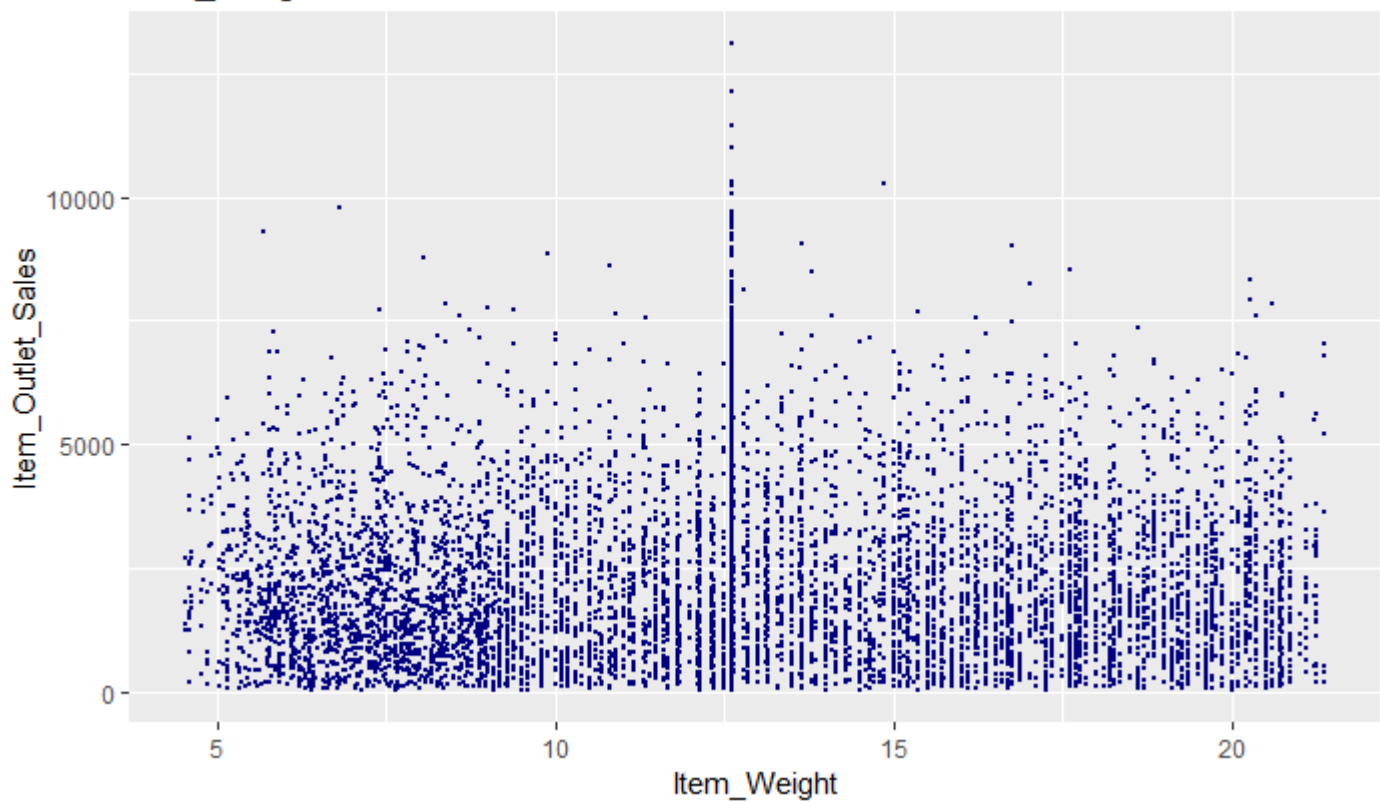


We have tried to visualise Item_Outlet_Sales with different Continuous Values:

Hide

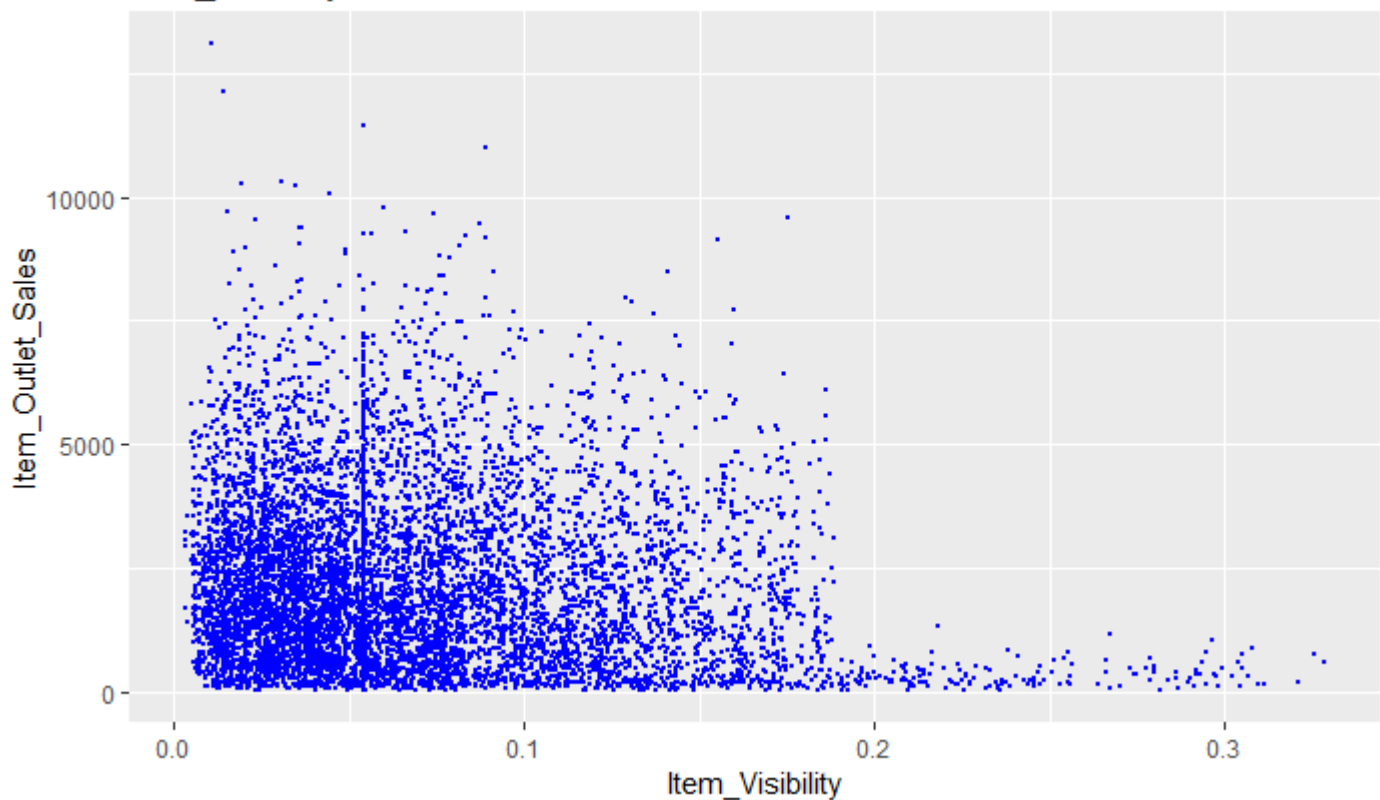
```
ggplot(new_train_combi, aes(Item_Weight, Item_Outlet_Sales)) + geom_point(size = .5, color="navy") + ggtitle("Item_Weight vs Item Outlet Sales")
```

Item_Weight vs Item Outlet Sales

[Hide](#)

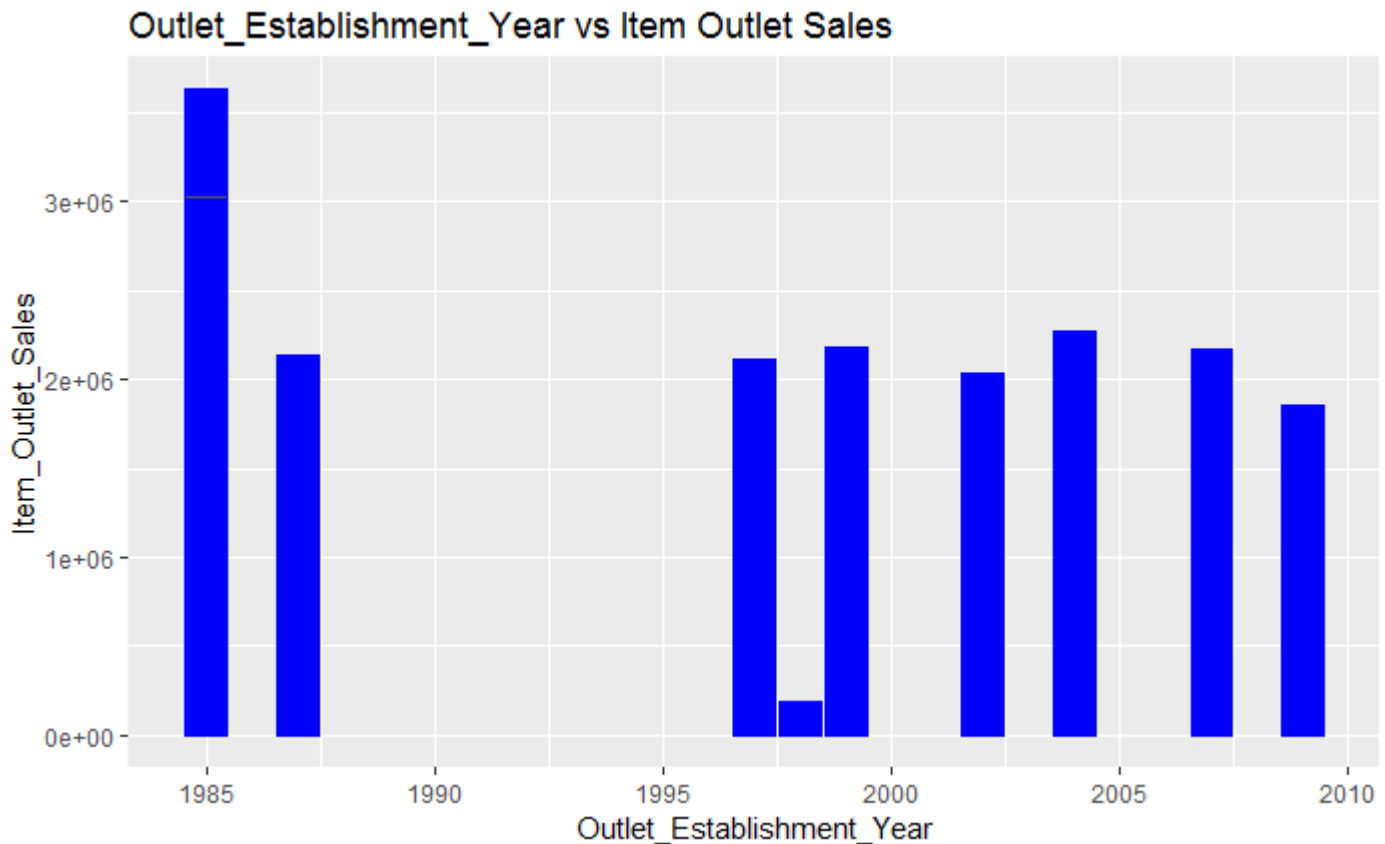
```
ggplot(new_train_combi, aes(Item_Visibility, Item_Outlet_Sales)) + geom_point(size = .5, color = "blue") + ggtitle("Item_Visibility vs Total Sales")
```

Item_Visibility vs Total Sales



Hide

```
ggplot(new_train_combi, aes(Outlet_Establishment_Year, Item_Outlet_Sales)) + geom_bar(stat = "id
entity", color="blue") + ggtitle("Outlet_Establishment_Year vs Item Outlet Sales")
```



Manipulating data with Label Encoding & Hot Encoding.

Creating a dataframe combi_encoded similar to combi.

Hide

```
combi_encoded=as.data.frame(combi)
str(combi_encoded)
```

Label Encoding. We will change the categorical variable Item_Fat_Content to numeric 0 & 1.

Hide

```
combi_encoded$Item_Fat_Content <- ifelse(combi_encoded$Item_Fat_Content == "Regular",1,0)
str(combi_encoded)
```

Hot Encoding. We will use dummy.data.frame() to split the catrgorical variable to a matrix of variables 0 and 1,

Hide

```
library(dummies)
combi_encoded_dummies <- dummy.data.frame(combi_encoded, names = c('Outlet_Size','Outlet_Location_Type','Outlet_Type'),sep = '_')

str(combi_encoded_dummies)
```

Now, We will save the dataframe with all columns with int and num and Drop the columns with Categorical variables/Factors.

Item_Identifier, Outlet_Identifier, Item_Type has not been converted to matrix because of the high no of factor levels, which we cannot compute due to limited system resources.

Hence we are removing the 3 variables from the final variable.

[Hide](#)

```
combi_encoded_dummies_drop <- select(combi_encoded_dummies, -c(Item_Identifier, Outlet_Identifier, Item_Type))
str(combi_encoded_dummies_drop)
```

```
'data.frame':  14204 obs. of  17 variables:
 $ Item_Weight          : num  9.3 5.92 17.5 19.2 8.93 ...
 $ Item_Fat_Content      : num  0 1 0 1 0 1 1 0 1 1 ...
 $ Item_Visibility       : num  0.016 0.0193 0.0168 0.054 0.054 ...
 $ Item_MRP              : num  249.8 48.3 141.6 182.1 53.9 ...
 $ Outlet_Establishment_Year : int  1999 2009 1999 1998 1987 2009 1987 1985 2002 2007 ...
 $ Outlet_Size_Other      : int  0 0 0 1 0 0 0 0 1 1 ...
 $ Outlet_Size_High       : int  0 0 0 0 1 0 1 0 0 0 ...
 $ Outlet_Size_Medium     : int  1 1 1 0 0 1 0 1 0 0 ...
 $ Outlet_Size_Small      : int  0 0 0 0 0 0 0 0 0 0 ...
 $ Outlet_Location_Type_Tier 1 : int  1 0 1 0 0 0 0 0 0 0 ...
 $ Outlet_Location_Type_Tier 2 : int  0 0 0 0 0 0 0 0 1 1 ...
 $ Outlet_Location_Type_Tier 3 : int  0 1 0 1 1 1 1 1 0 0 ...
 $ Outlet_Type_Grocery Store : int  0 0 0 1 0 0 0 0 0 0 ...
 $ Outlet_Type_Supermarket Type1: int  1 0 1 0 1 0 1 0 1 1 ...
 $ Outlet_Type_Supermarket Type2: int  0 1 0 0 0 1 0 0 0 0 ...
 $ Outlet_Type_Supermarket Type3: int  0 0 0 0 0 0 0 1 0 0 ...
 $ Item_Outlet_Sales      : num  3735 443 2097 732 995 ...
 - attr(*, "dummies")=List of 3
 ..$ Outlet_Size          : int  9 10 11 12
 ..$ Outlet_Location_Type : int  13 14 15
 ..$ Outlet_Type          : int  16 17 18 19
```

[Hide](#)

```
summary(combi_encoded_dummies_drop)
```

Item_Weight	Item_Fat_Content	Item_Visibility	Item_MRP	Outlet_Establishment_Year
Min. : 4.555	Min. : 0.0000	Min. : 0.003575	Min. : 31.29	Min. : 1985
1st Qu.: 9.300	1st Qu.: 0.0000	1st Qu.: 0.033143	1st Qu.: 94.01	1st Qu.: 1987
Median : 12.600	Median : 0.0000	Median : 0.054023	Median : 142.25	Median : 1999
Mean : 12.760	Mean : 0.3534	Mean : 0.069296	Mean : 141.00	Mean : 1998
3rd Qu.: 16.000	3rd Qu.: 1.0000	3rd Qu.: 0.094037	3rd Qu.: 185.86	3rd Qu.: 2004
Max. : 21.350	Max. : 1.0000	Max. : 0.328391	Max. : 266.89	Max. : 2009

Outlet_Size_Other	Outlet_Size_High	Outlet_Size_Medium	Outlet_Size_Small	Outlet_Location_Type_Tier 1
Min. : 0.0000	Min. : 0.0000	Min. : 0.0000	Min. : 0.0000	Min. : 0.0000
1st Qu.: 0.0000	1st Qu.: 0.0000	1st Qu.: 0.0000	1st Qu.: 0.0000	1st Qu.: 0.0000
Median : 0.0000	Median : 0.0000	Median : 0.0000	Median : 0.0000	Median : 0.0000
Mean : 0.2827	Mean : 0.1093	Mean : 0.3277	Mean : 0.2802	Mean : 0.2802
3rd Qu.: 1.0000	3rd Qu.: 0.0000	3rd Qu.: 1.0000	3rd Qu.: 1.0000	3rd Qu.: 1.0000
Max. : 1.0000	Max. : 1.0000	Max. : 1.0000	Max. : 1.0000	Max. : 1.0000

Outlet_Location_Type_Tier 2	Outlet_Location_Type_Tier 3	Outlet_Type_Grocery Store	Outlet_Type_Supermarket Type1
Min. : 0.0000	Min. : 0.0000	Min. : 0.0000	Min. : 0.0000
1st Qu.: 0.0000	1st Qu.: 0.0000	1st Qu.: 0.0000	1st Qu.: 0.0000
Median : 0.0000	Median : 0.0000	Median : 0.0000	Median : 1.0000
Mean : 0.3267	Mean : 0.3931	Mean : 0.1271	Mean : 0.6540
3rd Qu.: 1.0000	3rd Qu.: 1.0000	3rd Qu.: 0.0000	3rd Qu.: 1.0000
Max. : 1.0000	Max. : 1.0000	Max. : 1.0000	Max. : 1.0000

Outlet_Type_Supermarket Type2	Outlet_Type_Supermarket Type3	Item_Outlet_Sales
Min. : 0.0000	Min. : 0.0000	Min. : 1.0
1st Qu.: 0.0000	1st Qu.: 0.0000	1st Qu.: 1.0
Median : 0.0000	Median : 0.0000	Median : 559.3
Mean : 0.1088	Mean : 0.1098	Mean : 1309.3
3rd Qu.: 0.0000	3rd Qu.: 0.0000	3rd Qu.: 2163.2
Max. : 1.0000	Max. : 1.0000	Max. : 13087.0

Dividing data to Train & Test POST Label & Hot Encoding

[Hide](#)

```
new_train_combi_encoded_dummies_drop <- combi_encoded_dummies_drop %>% filter(Item_Outlet_Sales
!= 1)
new_test_combi_encoded_dummies_drop <- combi_encoded_dummies_drop %>% filter(Item_Outlet_Sales =
= 1)
str(new_train_combi_encoded_dummies_drop)
```

```
'data.frame': 8523 obs. of 17 variables:
 $ Item_Weight          : num  9.3 5.92 17.5 19.2 8.93 ...
 $ Item_Fat_Content      : num  0 1 0 1 0 1 1 0 1 1 ...
 $ Item_Visibility       : num  0.016 0.0193 0.0168 0.054 0.054 ...
 $ Item_MRP              : num  249.8 48.3 141.6 182.1 53.9 ...
 $ Outlet_Establishment_Year : int  1999 2009 1999 1998 1987 2009 1987 1985 2002 2007 ...
 $ Outlet_Size_Other     : int  0 0 0 1 0 0 0 0 1 1 ...
 $ Outlet_Size_High      : int  0 0 0 0 1 0 1 0 0 0 ...
 $ Outlet_Size_Medium    : int  1 1 1 0 0 1 0 1 0 0 ...
 $ Outlet_Size_Small     : int  0 0 0 0 0 0 0 0 0 0 ...
 $ Outlet_Location_Type_Tier 1 : int  1 0 1 0 0 0 0 0 0 0 ...
 $ Outlet_Location_Type_Tier 2 : int  0 0 0 0 0 0 0 0 1 1 ...
 $ Outlet_Location_Type_Tier 3 : int  0 1 0 1 1 1 1 1 0 0 ...
 $ Outlet_Type_Grocery Store : int  0 0 0 1 0 0 0 0 0 0 ...
 $ Outlet_Type_Supermarket Type1: int  1 0 1 0 1 0 1 0 1 1 ...
 $ Outlet_Type_Supermarket Type2: int  0 1 0 0 0 1 0 0 0 0 ...
 $ Outlet_Type_Supermarket Type3: int  0 0 0 0 0 0 0 1 0 0 ...
 $ Item_Outlet_Sales      : num  3735 443 2097 732 995 ...
- attr(*, "dummies")=List of 3
 ..$ Outlet_Size          : int  9 10 11 12
 ..$ Outlet_Location_Type: int  13 14 15
 ..$ Outlet_Type          : int  16 17 18 19
```

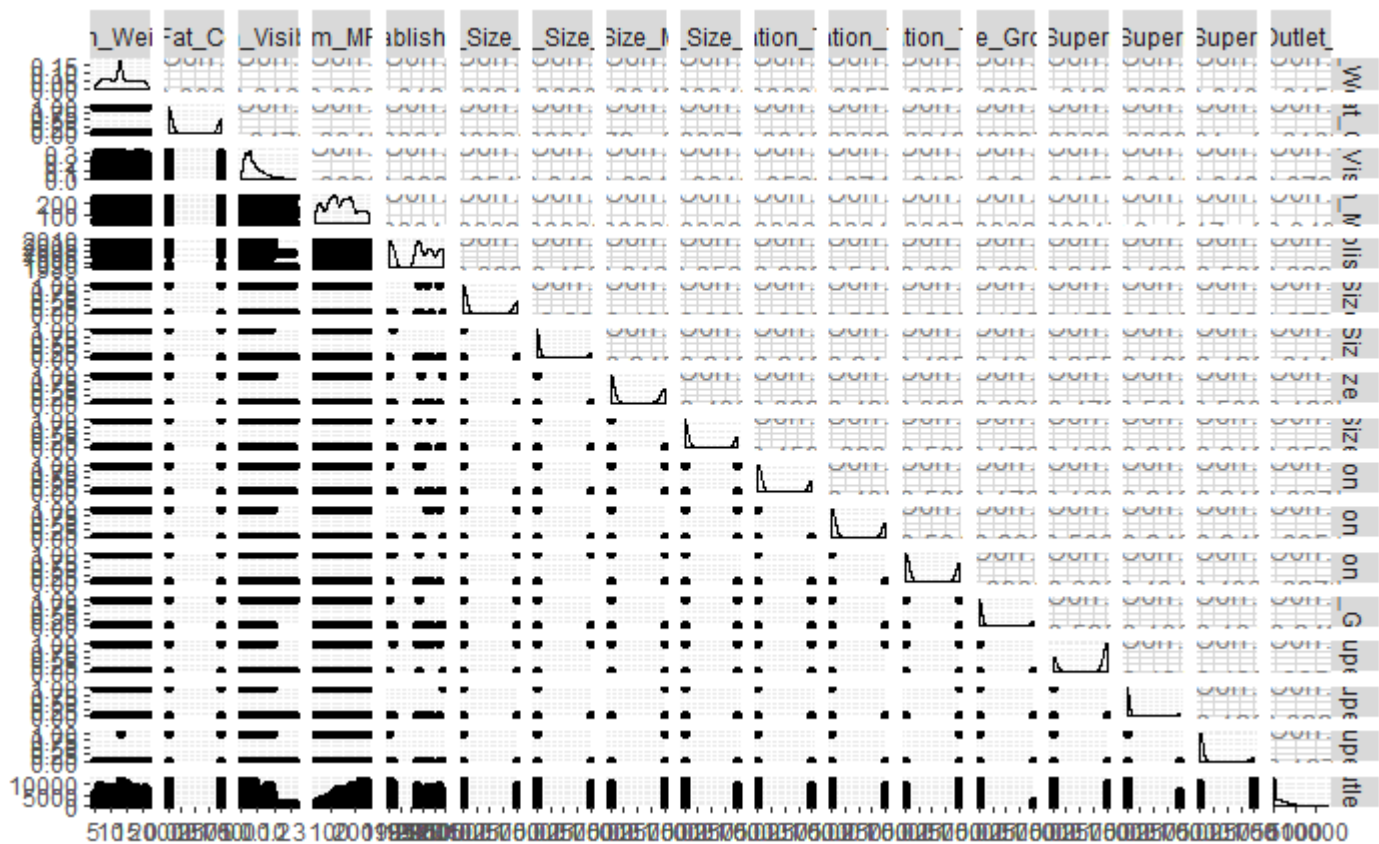
Hide

```
str(new_test_combi_encoded_dummies_drop)
```

```
'data.frame': 5681 obs. of 17 variables:
 $ Item_Weight          : num  20.75 8.3 14.6 7.32 12.6 ...
 $ Item_Fat_Content      : num  0 1 0 0 1 1 1 0 1 0 ...
 $ Item_Visibility       : num  0.00756 0.03843 0.09957 0.01539 0.1186 ...
 $ Item_MRP              : num  107.9 87.3 241.8 155 234.2 ...
 $ Outlet_Establishment_Year : int  1999 2007 1998 2007 1985 1997 2009 1985 2002 2007 ...
 $ Outlet_Size_Other     : int  0 1 1 1 0 0 0 0 1 1 ...
 $ Outlet_Size_High      : int  0 0 0 0 0 0 0 0 0 0 ...
 $ Outlet_Size_Medium    : int  1 0 0 0 1 0 1 1 0 0 ...
 $ Outlet_Size_Small     : int  0 0 0 0 0 1 0 0 0 0 ...
 $ Outlet_Location_Type_Tier 1 : int  1 0 0 0 0 1 0 0 0 0 ...
 $ Outlet_Location_Type_Tier 2 : int  0 1 0 1 0 0 0 0 1 1 ...
 $ Outlet_Location_Type_Tier 3 : int  0 0 1 0 1 0 1 1 0 0 ...
 $ Outlet_Type_Grocery Store : int  0 0 1 0 0 0 0 0 0 0 ...
 $ Outlet_Type_Supermarket Type1: int  1 1 0 1 0 1 0 0 1 1 ...
 $ Outlet_Type_Supermarket Type2: int  0 0 0 0 0 0 1 0 0 0 ...
 $ Outlet_Type_Supermarket Type3: int  0 0 0 0 1 0 0 1 0 0 ...
 $ Item_Outlet_Sales      : num  1 1 1 1 1 1 1 1 1 1 ...
- attr(*, "dummies")=List of 3
 ..$ Outlet_Size          : int  9 10 11 12
 ..$ Outlet_Location_Type: int  13 14 15
 ..$ Outlet_Type          : int  16 17 18 19
```

Hide

```
library(GGally)
ggpairs(combi_encoded_dummies_drop)
```



Linear (Multiple) Regression

Amount of correlation present in our predictor variables

Hide

```
cor(new_train_combi_encoded_dummies_drop)
```

	Item_Weight	Item_Fat_Content	Item_Visibility	Item_MRP
Item_Weight	1.000000000	-0.0210920104	-0.018772588	0.0249505601
Item_Fat_Content	-0.021092010	1.000000000	0.049793752	0.0060628994
Item_Visibility	-0.018772588	0.049793752	1.000000000	-0.0045367831
Item_MRP	0.024950560	0.0060628994	-0.004536783	1.000000000
Outlet_Establishment_Year	0.007739014	0.0031506634	-0.078272866	0.0050199162
Outlet_Size_Other	-0.005190112	-0.0010847801	0.051641561	-0.0067540030
Outlet_Size_High	0.015976179	-0.0021320419	-0.043643327	0.0024375785
Outlet_Size_Medium	-0.002790703	0.0046714179	-0.083994282	-0.0045100820
Outlet_Size_Small	-0.002980744	-0.0023130394	0.066331288	0.0097927692
Outlet_Location_Type_Tier 1	0.002083178	0.0031548795	0.063767534	-0.0012290862
Outlet_Location_Type_Tier 2	-0.007382332	-0.0032717388	-0.073390251	0.0019513073
Outlet_Location_Type_Tier 3	0.005173738	0.0002410231	0.011843986	-0.0007437168
Outlet_Type_Grocery Store	-0.004778323	-0.0029242966	0.299204770	-0.0042771353
Outlet_Type_Supermarket Type1	0.011747051	0.0005332351	-0.152830781	0.0048854837
Outlet_Type_Supermarket Type2	0.004955601	0.0021294095	-0.033373806	0.0038499211
Outlet_Type_Supermarket Type3	-0.017723313	0.0001825365	-0.053023468	-0.0067136994
Item_Outlet_Sales	0.009692876	0.0187185336	-0.134097091	0.5675744467
	Outlet_Establishment_Year	Outlet_Size_Other	Outlet_Size_High	Outlet_Size_Medium
Item_Weight	0.007739014	-0.005190112	0.015976179	-0.002790703
Item_Fat_Content	0.003150663	-0.001084780	-0.002132042	0.004671418
Item_Visibility	-0.078272866	0.051641561	-0.043643327	-0.083994282
Item_MRP	0.005019916	-0.006754003	0.002437579	-0.004510082
Outlet_Establishment_Year	1.000000000	0.387635656	-0.453388454	-0.016345705
Outlet_Size_Other	0.387635656	1.000000000	-0.220008664	-0.438368642
Outlet_Size_High	-0.453388454	-0.220008664	1.000000000	-0.244633888
Outlet_Size_Medium	-0.016345705	-0.438368642	-0.244633888	1.000000000
Outlet_Size_Small	-0.056566813	-0.391733940	-0.218609151	-0.435580104
Outlet_Location_Type_Tier 1	-0.201690130	-0.391733940	-0.218609151	0.082072274
Outlet_Location_Type_Tier 2	0.540819608	0.592969531	-0.244112933	-0.486396549
Outlet_Location_Type_Tier 3	-0.333894725	-0.209236546	0.435418920	0.391616506
Outlet_Type_Grocery Store	-0.281195730	0.194602128	-0.133686090	-0.266370372
Outlet_Type_Supermarket Type1	0.245069762	0.152307648	0.254668077	-0.471782330
Outlet_Type_Supermarket Type2	0.466336465	-0.219478216	-0.122480954	0.500670430
Outlet_Type_Supermarket Type3	-0.538072347	-0.220406028	-0.122998724	0.502786939
Item_Outlet_Sales	-0.049134970	-0.131973256	0.024170053	

0.204701320			
	Outlet_Size_Small	Outlet_Location_Type_Tier 1	Outlet_Location_Type
_Tier 2			
Item_Weight	-0.002980744	0.002083178	-0.00
7382332			
Item_Fat_Content	-0.002313039	0.003154879	-0.00
3271739			
Item_Visibility	0.066331288	0.063767534	-0.07
3390251			
Item_MRP	0.009792769	-0.001229086	0.00
1951307			
Outlet_Establishment_Year	-0.056566813	-0.201690130	0.54
0819608			
Outlet_Size_Other	-0.391733940	-0.391733940	0.59
2969531			
Outlet_Size_High	-0.218609151	-0.218609151	-0.24
4112933			
Outlet_Size_Medium	-0.435580104	0.082072274	-0.48
6396549			
Outlet_Size_Small	1.000000000	0.458963522	0.08
3381305			
Outlet_Location_Type_Tier 1	0.458963522	1.000000000	-0.43
4652524			
Outlet_Location_Type_Tier 2	0.083381305	-0.434652524	1.00
0000000			
Outlet_Location_Type_Tier 3	-0.502066266	-0.502066266	-0.56
0639241			
Outlet_Type_Grocery Store	0.176158327	0.176158327	-0.26
5803129			
Outlet_Type_Supermarket Type1	0.163388083	0.163388083	0.50
6347158			
Outlet_Type_Supermarket Type2	-0.218082078	-0.218082078	-0.24
3524369			
Outlet_Type_Supermarket Type3	-0.219003987	-0.219003987	-0.24
4553832			
Item_Outlet_Sales	-0.098402699	-0.111287125	0.05
8261357			
	Outlet_Location_Type_Tier 3	Outlet_Type_Grocery Store	Outlet_Type_
Supermarket Type1			
Item_Weight	0.0051737383	-0.004778323	
0.0117470508			
Item_Fat_Content	0.0002410231	-0.002924297	
0.0005332351			
Item_Visibility	0.0118439862	0.299204770	
-0.1528307806			
Item_MRP	-0.0007437168	-0.004277135	
0.0048854837			
Outlet_Establishment_Year	-0.3338947248	-0.281195730	
0.2450697621			
Outlet_Size_Other	-0.2092365463	0.194602128	
0.1523076482			
Outlet_Size_High	0.4354189199	-0.133686090	
0.2546680773			
Outlet_Size_Medium	0.3916165060	-0.266370372	

-0.4717823295			
Outlet_Size_Small	-0.5020662661	0.176158327	
0.1633880834			
Outlet_Location_Type_Tier 1	-0.5020662661	0.176158327	
0.1633880834			
Outlet_Location_Type_Tier 2	-0.5606392409	-0.265803129	
0.5063471581			
Outlet_Location_Type_Tier 3	1.0000000000	0.093276443	
-0.6364646571			
Outlet_Type_Grocery Store	0.0932764434	1.0000000000	
-0.5249424714			
Outlet_Type_Supermarket Type1	-0.6364646571	-0.524942471	
1.0000000000			
Outlet_Type_Supermarket Type2	0.4343691117	-0.133363769	
-0.4809434894			
Outlet_Type_Supermarket Type3	0.4362053421	-0.133927544	
-0.4829766060			
Item_Outlet_Sales	0.0463761913	-0.411727080	
0.1087652555			
	Outlet_Type_Supermarket Type2	Outlet_Type_Supermarket Type3	Item_O
utlet_Sales			
Item_Weight	0.004955601	-0.0177233134	
0.009692876			
Item_Fat_Content	0.002129410	0.0001825365	
0.018718534			
Item_Visibility	-0.033373806	-0.0530234675	-
0.134097091			
Item_MRP	0.003849921	-0.0067136994	
0.567574447			
Outlet_Establishment_Year	0.466336465	-0.5380723466	-
0.049134970			
Outlet_Size_Other	-0.219478216	-0.2204060278	-
0.131973256			
Outlet_Size_High	-0.122480954	-0.1229987236	
0.024170053			
Outlet_Size_Medium	0.500670430	0.5027869391	
0.204701320			
Outlet_Size_Small	-0.218082078	-0.2190039874	-
0.098402699			
Outlet_Location_Type_Tier 1	-0.218082078	-0.2190039874	-
0.111287125			
Outlet_Location_Type_Tier 2	-0.243524369	-0.2445538319	
0.058261357			
Outlet_Location_Type_Tier 3	0.434369112	0.4362053421	
0.046376191			
Outlet_Type_Grocery Store	-0.133363769	-0.1339275441	-
0.411727080			
Outlet_Type_Supermarket Type1	-0.480943489	-0.4829766060	
0.108765256			
Outlet_Type_Supermarket Type2	1.0000000000	-0.1227021700	-
0.038058540			
Outlet_Type_Supermarket Type3	-0.122702170	1.0000000000	
0.311192046			

Item_Outlet_Sales
1.000000000

-0.038058540

0.3111920462

Hide

summary(linear_model)

Call:

```
lm(formula = Item_Outlet_Sales ~ ., data = new_train_combi_encoded_dummies_drop)
```

Residuals:

Min	1Q	Median	3Q	Max
-4313.0	-675.6	-87.6	571.3	7917.6

Coefficients: (3 not defined because of singularities)

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	-6.666e+04	2.078e+04	-3.207	0.00134	**
Item_Weight	-6.276e-01	2.895e+00	-0.217	0.82837	
Item_Fat_Content	5.151e+01	2.562e+01	2.010	0.04443	*
Item_Visibility	-2.410e+02	2.626e+02	-0.918	0.35880	
Item_MRP	1.556e+01	1.964e-01	79.232	< 2e-16	***
Outlet_Establishment_Year	3.433e+01	1.048e+01	3.276	0.00106	**
Outlet_Size_Other	-1.440e+02	4.564e+01	-3.156	0.00161	**
Outlet_Size_High	6.953e+02	2.552e+02	2.725	0.00645	**
Outlet_Size_Medium	2.911e+01	5.637e+01	0.516	0.60556	
Outlet_Size_Small	NA	NA	NA	NA	
`Outlet_Location_Type_Tier 1`	3.207e+02	1.547e+02	2.073	0.03824	*
`Outlet_Location_Type_Tier 2`	2.253e+02	1.005e+02	2.242	0.02496	*
`Outlet_Location_Type_Tier 3`	NA	NA	NA	NA	
`Outlet_Type_Grocery Store`	-3.633e+03	1.779e+02	-20.415	< 2e-16	***
`Outlet_Type_Supermarket Type1`	-2.154e+03	2.944e+02	-7.319	2.73e-13	***
`Outlet_Type_Supermarket Type2`	-2.551e+03	2.569e+02	-9.930	< 2e-16	***
`Outlet_Type_Supermarket Type3`	NA	NA	NA	NA	

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1128 on 8509 degrees of freedom

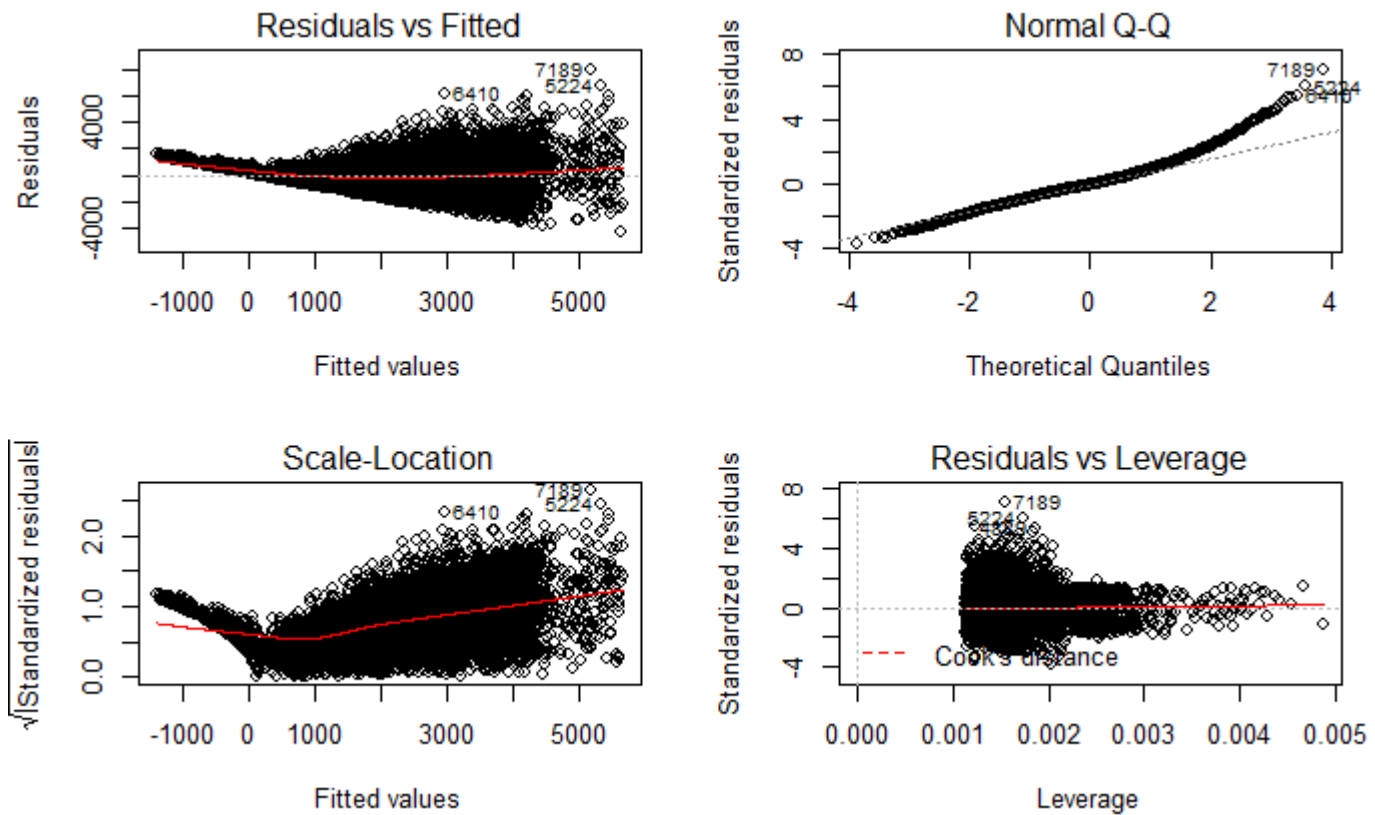
Multiple R-squared: 0.5635, Adjusted R-squared: 0.5628

F-statistic: 844.8 on 13 and 8509 DF, p-value: < 2.2e-16

Using Regression plot:

Hide

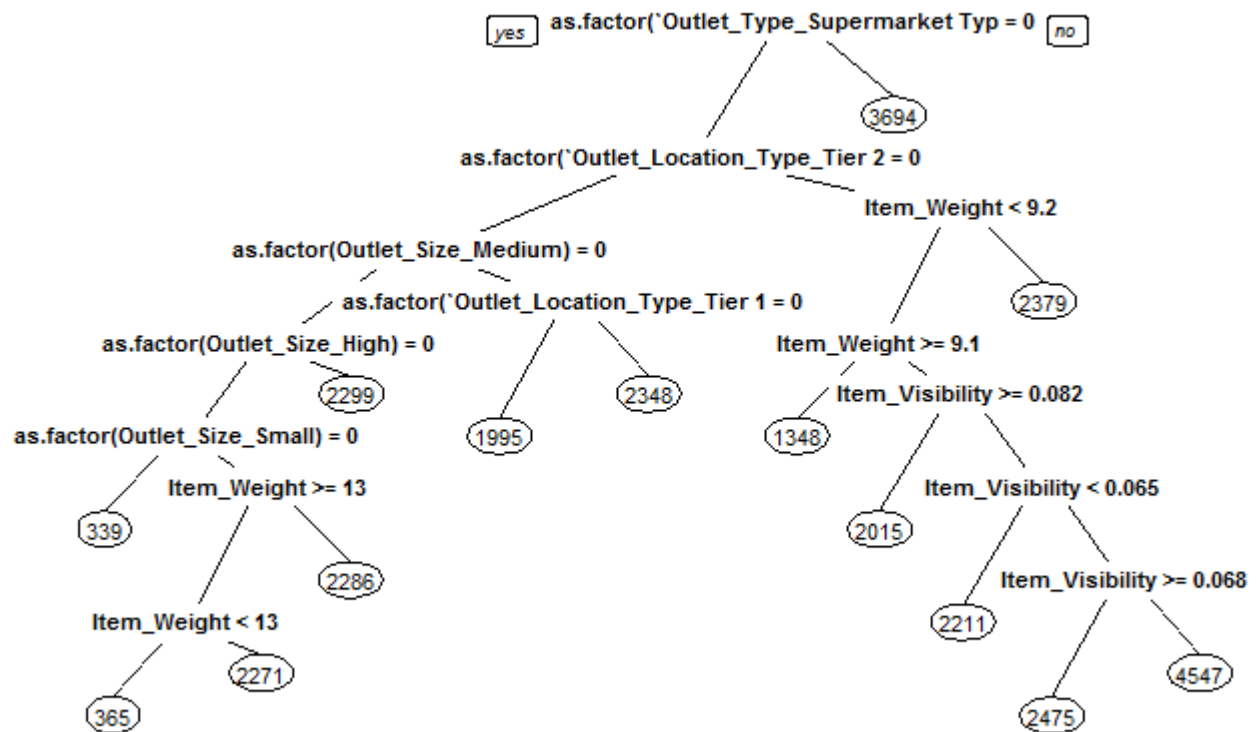
```
par(mfrow=c(2,2))
plot(linear_model)
```



Random Forest

[Hide](#)

```
#load randomForest library
library(randomForest)
library(rpart)
library(e1071)
library(rpart.plot)
library(caret)
formula_tree <- as.formula(Item_Outlet_Sales ~ Item_Weight +
Item_Visibility +
as.factor(Outlet_Size_High) +
as.factor(Outlet_Size_Medium) +
as.factor(Outlet_Size_Small) +
as.factor(`Outlet_Location_Type_Tier 1`) +
as.factor(`Outlet_Location_Type_Tier 2`) +
as.factor(`Outlet_Location_Type_Tier 3`) +
as.factor(`Outlet_Type_Supermarket Type3`))
tree1 <- rpart(formula_tree, data = new_train_combi_encoded_dummies_drop, control = rpart.control(cp=0.001))
prp(tree1)
```



Hide

summary(tree1)

Call:

```
rpart(formula = formula_tree, data = new_train_combi_encoded_dummies_drop,
      control = rpart.control(cp = 0.001))
n= 8523
```

	CP	nsplit	rel error	xerror	xstd
1	0.096840490	0	1.0000000	1.0002935	0.02059661
2	0.021597713	1	0.9031595	0.9037103	0.01738674
3	0.002330376	7	0.7636181	0.7651574	0.01617594
4	0.001007738	8	0.7612877	0.7650304	0.01618859
5	0.001000000	13	0.7562490	0.7712937	0.01629126

Variable importance

	Item_Weight
as.factor(`Outlet_Type_Supermarket Type3`)	33
as.factor(Outlet_Size_Small)	13
as.factor(`Outlet_Location_Type_Tier 2`)	6
as.factor(Outlet_Size_Medium)	6
Item_Visibility	4
as.factor(Outlet_Size_High)	17
as.factor(`Outlet_Location_Type_Tier 1`)	10
as.factor(`Outlet_Location_Type_Tier 3`)	6
	5

Node number 1: 8523 observations, complexity param=0.09684049

mean=2181.289, MSE=2911799

left son=2 (7588 obs) right son=3 (935 obs)

Primary splits:

```
as.factor(`Outlet_Type_Supermarket Type3`) splits as LR, improve=0.096840490, (0 missing)
as.factor(Outlet_Size_Medium) splits as LR, improve=0.041902630, (0 missing)
Item_Visibility < 0.1876999 to the right, improve=0.02274282
0, (0 missing)
as.factor(`Outlet_Location_Type_Tier 1`) splits as RL, improve=0.012384820, (0 missing)
as.factor(Outlet_Size_Small) splits as RL, improve=0.009683091, (0 missing)
```

Node number 2: 7588 observations, complexity param=0.02159771

mean=1994.887, MSE=2396599

left son=4 (4803 obs) right son=5 (2785 obs)

Primary splits:

```
as.factor(`Outlet_Location_Type_Tier 2`) splits as LR, improve=0.026204960, (0 missing)
Item_Visibility < 0.1876999 to the right, improve=0.025033390,
(0 missing)
as.factor(`Outlet_Location_Type_Tier 3`) splits as RL, improve=0.013460230, (0 missing)
Item_Weight < 12.625 to the left, improve=0.008887997,
(0 missing)
as.factor(Outlet_Size_High) splits as LR, improve=0.005403372, (0 missing)
Surrogate splits:
as.factor(`Outlet_Location_Type_Tier 3`) splits as RL, agree=0.685, adj=0.143, (0 split)
as.factor(`Outlet_Location_Type_Tier 1`) splits as RL, agree=0.682, adj=0.133, (0 split)
```

Node number 3: 935 observations

mean=3694.039, MSE=4522521

```

Node number 4: 4803 observations,    complexity param=0.02159771
mean=1804.057, MSE=2346885
left son=8 (2945 obs) right son=9 (1858 obs)
Primary splits:
  as.factor(Outlet_Size_Medium) splits as LR, improve=0.036417060, (0 missing)
  Item_Visibility               < 0.1871346 to the right, improve=0.032668820, (0 missing)
  as.factor(Outlet_Size_High)   splits as LR, improve=0.025130610, (0 missing)
  Item_Weight                   < 12.725 to the left, improve=0.014730420, (0 missing)
  as.factor(Outlet_Size_Small) splits as RL, improve=0.009643201, (0 missing)
Surrogate splits:
  as.factor(Outlet_Size_Small) splits as RL, agree=0.690, adj=0.200, (0 split)
  Item_Visibility               < 0.004418756 to the right, agree=0.613, adj=0.001, (0 split)

Node number 5: 2785 observations,    complexity param=0.001007738
mean=2323.991, MSE=2311222
left son=10 (816 obs) right son=11 (1969 obs)
Primary splits:
  Item_Weight                   < 9.24 to the left, improve=0.003134485, (0 missing)
  as.factor(Outlet_Size_Small) splits as LR, improve=0.002861338, (0 missing)
  Item_Visibility               < 0.02369722 to the right, improve=0.002025882, (0 missing)
Surrogate splits:
  Item_Visibility < 0.006036989 to the left, agree=0.71, adj=0.009, (0 split)

Node number 8: 2945 observations,    complexity param=0.02159771
mean=1571.848, MSE=2350193
left son=16 (2013 obs) right son=17 (932 obs)
Primary splits:
  as.factor(Outlet_Size_High)   splits as LR, improve=1.041629e-01, (0 missing)
  Item_Visibility               < 0.1478574 to the right, improve=4.266678e-02,
(0 missing)
  Item_Weight                   < 12.625 to the left, improve=2.770336e-02,
(0 missing)
  as.factor(Outlet_Size_Small)   splits as LR, improve=7.875173e-06, (0 missing)
  as.factor(`Outlet_Location_Type_Tier 1`) splits as LR, improve=7.875173e-06, (0 missing)
Surrogate splits:
  as.factor(Outlet_Size_Small)   splits as RL, agree=0.812, adj=0.405, (0 split)
  as.factor(`Outlet_Location_Type_Tier 1`) splits as RL, agree=0.812, adj=0.405, (0 split)
  as.factor(`Outlet_Location_Type_Tier 3`) splits as LR, agree=0.812, adj=0.405, (0 split)
  Item_Visibility               < 0.008760024 to the right, agree=0.687, adj=0.01
2, (0 split)

Node number 9: 1858 observations,    complexity param=0.002330376
mean=2172.117, MSE=2120707
left son=18 (928 obs) right son=19 (930 obs)
Primary splits:
  as.factor(`Outlet_Location_Type_Tier 1`) splits as LR, improve=0.0146775500, (0 missing)
  as.factor(`Outlet_Location_Type_Tier 3`) splits as RL, improve=0.0146775500, (0 missing)
  Item_Visibility               < 0.1599065 to the right, improve=0.0021485360,
(0 missing)
  Item_Weight                   < 21.225 to the left, improve=0.0007303589,
(0 missing)
Surrogate splits:
  Item_Weight < 8.305 to the left, agree=0.513, adj=0.026, (0 split)
  Item_Visibility < 0.0418986 to the right, agree=0.512, adj=0.023, (0 split)

```

Node number 10: 816 observations, complexity param=0.001007738
 mean=2191.775, MSE=1960975
 left son=20 (25 obs) right son=21 (791 obs)
 Primary splits:
 Item_Weight < 9.1025 to the right, improve=0.011483350, (0 missing)
 Item_Visibility < 0.0819964 to the right, improve=0.008333672, (0 missing)
 as.factor(Outlet_Size_Small) splits as LR, improve=0.003297263, (0 missing)

Node number 11: 1969 observations
 mean=2378.784, MSE=2446127

Node number 16: 2013 observations, complexity param=0.02159771
 mean=1235.186, MSE=1992511
 left son=32 (555 obs) right son=33 (1458 obs)
 Primary splits:
 as.factor(Outlet_Size_Small) splits as LR, improve=0.15331690, (0 missing)
 as.factor(`Outlet_Location_Type_Tier 1`) splits as LR, improve=0.15331690, (0 missing)
 as.factor(`Outlet_Location_Type_Tier 3`) splits as RL, improve=0.15331690, (0 missing)
 Item_Visibility < 0.1053019 to the right, improve=0.04639051,
 (0 missing)
 Item_Weight < 12.55 to the right, improve=0.02963642,
 (0 missing)
 Surrogate splits:
 Item_Weight < 21.3 to the right, agree=0.725, adj=0.002, (0 split)

Node number 17: 932 observations
 mean=2298.995, MSE=2349196

Node number 18: 928 observations
 mean=1995.499, MSE=1891151

Node number 19: 930 observations
 mean=2348.355, MSE=2287583

Node number 20: 25 observations
 mean=1347.686, MSE=467144.2

Node number 21: 791 observations, complexity param=0.001007738
 mean=2218.453, MSE=1984958
 left son=42 (245 obs) right son=43 (546 obs)
 Primary splits:
 Item_Visibility < 0.0819964 to the right, improve=0.009339677, (0 missing)
 Item_Weight < 5.4625 to the right, improve=0.004727454, (0 missing)
 as.factor(Outlet_Size_Small) splits as LR, improve=0.003139342, (0 missing)
 Surrogate splits:
 Item_Weight < 4.795 to the left, agree=0.702, adj=0.037, (0 split)

Node number 32: 555 observations
 mean=339.3517, MSE=73316.71

Node number 33: 1458 observations, complexity param=0.02159771
 mean=1576.193, MSE=2301297
 left son=66 (999 obs) right son=67 (459 obs)

Primary splits:

Item_Weight < 12.55 to the right, improve=0.1005474, (0 missing)

Item_Visibility < 0.1475278 to the right, improve=0.0502919, (0 missing)

Surrogate splits:

Item_Visibility < 0.01167985 to the right, agree=0.69, adj=0.015, (0 split)

Node number 42: 245 observations

mean=2015.192, MSE=1690564

Node number 43: 546 observations, complexity param=0.001007738

mean=2309.66, MSE=2090200

left son=86 (444 obs) right son=87 (102 obs)

Primary splits:

Item_Visibility < 0.06471562 to the left, improve=0.020248240, (0 missing)

Item_Weight < 6.105 to the right, improve=0.003617898, (0 missing)

as.factor(Outlet_Size_Small) splits as LR, improve=0.002980943, (0 missing)

Surrogate splits:

Item_Weight < 9.05 to the left, agree=0.815, adj=0.01, (0 split)

Node number 66: 999 observations, complexity param=0.02159771

mean=1250.134, MSE=1991107

left son=132 (535 obs) right son=133 (464 obs)

Primary splits:

Item_Weight < 12.625 to the left, improve=0.45383950, (0 missing)

Item_Visibility < 0.1054815 to the right, improve=0.05700259, (0 missing)

Surrogate splits:

Item_Visibility < 0.05747654 to the right, agree=0.64, adj=0.224, (0 split)

Node number 67: 459 observations

mean=2285.85, MSE=2241414

Node number 86: 444 observations

mean=2211.056, MSE=1716776

Node number 87: 102 observations, complexity param=0.001007738

mean=2738.879, MSE=3489142

left son=174 (89 obs) right son=175 (13 obs)

Primary splits:

Item_Visibility < 0.06799998 to the right, improve=0.13690350, (0 missing)

Item_Weight < 5.7575 to the right, improve=0.02053833, (0 missing)

as.factor(Outlet_Size_Small) splits as LR, improve=0.01802642, (0 missing)

Node number 132: 535 observations

mean=364.8547, MSE=131842.8

Node number 133: 464 observations

mean=2270.877, MSE=2189312

Node number 174: 89 observations

mean=2474.734, MSE=2568266

Node number 175: 13 observations

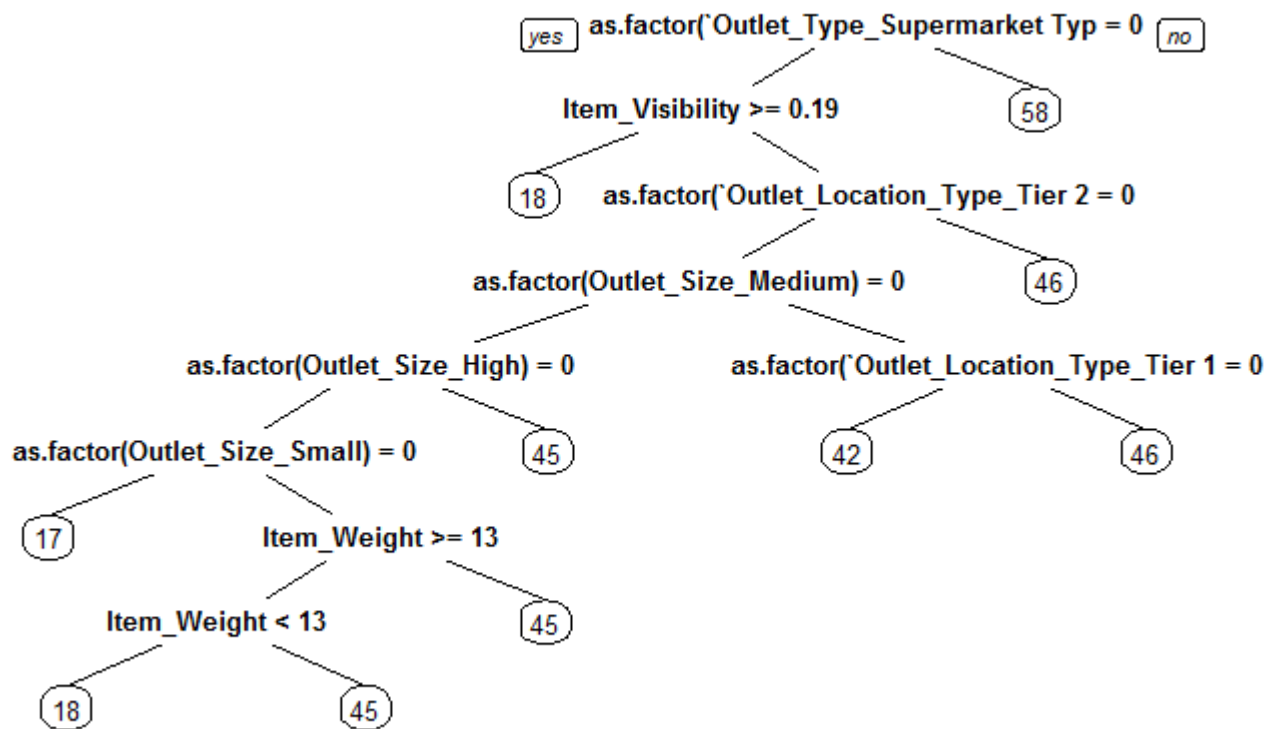
mean=4547.26, MSE=6045685

Hide

```

formula_sqrt_tree <- as.formula(sqrt(Item_Outlet_Sales) ~ Item_Weight +
Item_Visibility +
as.factor(Outlet_Size_High) +
as.factor(Outlet_Size_Medium) +
as.factor(Outlet_Size_Small) +
as.factor(`Outlet_Location_Type_Tier 1`) +
as.factor(`Outlet_Location_Type_Tier 2`) +
as.factor(`Outlet_Location_Type_Tier 3`) +
as.factor(`Outlet_Type_Supermarket Type3`))
tree2 <- rpart(formula_sqrt_tree, data = new_train_combi_encoded_dummies_drop, control = rpart.c
ontrol(cp=0.001))
prp(tree2)

```



Hide

```
summary(tree2)
```


Call:

```
rpart(formula = formula_sqrt_tree, data = new_train_combi_encoded_dummies_drop,
      control = rpart.control(cp = 0.001))
n= 8523
```

	CP	nsplit	rel error	xerror	xstd
1	0.084734551	0	1.0000000	1.0000587	0.013488517
2	0.033643184	1	0.9152654	0.9155306	0.011935872
3	0.002451204	8	0.6670346	0.6681327	0.009953933
4	0.001000000	9	0.6645834	0.6658112	0.009917513

Variable importance

	Item_Weight	Item_Visibility
as.factor(`Outlet_Type_Supermarket Type3`)	20	19
as.factor(Outlet_Size_Small)	17	11
as.factor(Outlet_Size_High)	10	6
as.factor(`Outlet_Location_Type_Tier 2`)	6	6
as.factor(`Outlet_Location_Type_Tier 3`)	6	

Node number 1: 8523 observations, complexity param=0.08473455

mean=42.94478, MSE=337.035

left son=2 (7588 obs) right son=3 (935 obs)

Primary splits:

as.factor(`Outlet_Type_Supermarket Type3`) splits as LR, improve=0.08473455, (0 missing)
 as.factor(Outlet_Size_Medium) splits as LR, improve=0.04830326, (0 missing)
 Item_Visibility < 0.1876999 to the right, improve=0.03812571,
 (0 missing)
 as.factor(`Outlet_Location_Type_Tier 1`) splits as RL, improve=0.01521102, (0 missing)
 as.factor(Outlet_Size_Small) splits as RL, improve=0.01226258, (0 missing)

Node number 2: 7588 observations, complexity param=0.03364318

mean=41.06888, MSE=308.234

left son=4 (169 obs) right son=5 (7419 obs)

Primary splits:

Item_Visibility < 0.1876999 to the right, improve=0.040214080,
 (0 missing)
 as.factor(`Outlet_Location_Type_Tier 2`) splits as LR, improve=0.039407540, (0 missing)
 as.factor(`Outlet_Location_Type_Tier 3`) splits as RL, improve=0.019028330, (0 missing)
 Item_Weight < 12.625 to the left, improve=0.012605810,
 (0 missing)
 as.factor(Outlet_Size_Medium) splits as LR, improve=0.008860972, (0 missing)

Node number 3: 935 observations

mean=58.16867, MSE=310.4447

Node number 4: 169 observations

mean=17.74189, MSE=57.2304

Node number 5: 7419 observations, complexity param=0.03364318

```

mean=41.60025, MSE=301.274
left son=10 (4636 obs) right son=11 (2783 obs)
Primary splits:
  as.factor(`Outlet_Location_Type_Tier 2`) splits as LR, improve=0.032631850, (0 missing)
  as.factor(`Outlet_Location_Type_Tier 3`) splits as RL, improve=0.017205290, (0 missing)
  Item_Weight < 12.625 to the left, improve=0.009803359,
(0 missing)
  Item_Visibility < 0.09709373 to the right, improve=0.008435584,
(0 missing)
  as.factor(Outlet_Size_Medium) splits as LR, improve=0.006199232, (0 missing)
Surrogate splits:
  as.factor(`Outlet_Location_Type_Tier 3`) splits as RL, agree=0.690, adj=0.173, (0 split)
  as.factor(`Outlet_Location_Type_Tier 1`) splits as RL, agree=0.685, adj=0.161, (0 split)
  as.factor(Outlet_Size_Medium) splits as RL, agree=0.625, adj=0.001, (0 split)
  Item_Weight < 4.795 to the right, agree=0.625, adj=0.00
0, (0 split)

Node number 10: 4636 observations, complexity param=0.03364318
mean=39.17092, MSE=322.0103
left son=20 (2780 obs) right son=21 (1856 obs)
Primary splits:
  as.factor(Outlet_Size_Medium) splits as LR, improve=0.04767614, (0 missing)
  as.factor(Outlet_Size_High) splits as LR, improve=0.02812387, (0 missing)
  Item_Weight < 12.625 to the left, improve=0.01683184, (0 missing)
  Item_Visibility < 0.09722637 to the right, improve=0.01237150, (0 missing)
  as.factor(Outlet_Size_Small) splits as RL, improve=0.01158969, (0 missing)
Surrogate splits:
  as.factor(Outlet_Size_Small) splits as RL, agree=0.697, adj=0.242, (0 split)
  as.factor(Outlet_Size_High) splits as RL, agree=0.601, adj=0.004, (0 split)
  Item_Visibility < 0.004418756 to the right, agree=0.600, adj=0.001, (0 split)

Node number 11: 2783 observations
mean=45.6471, MSE=240.5229

Node number 20: 2780 observations, complexity param=0.03364318
mean=35.96943, MSE=351.9025
left son=40 (1848 obs) right son=41 (932 obs)
Primary splits:
  as.factor(Outlet_Size_High) splits as LR, improve=0.1213224000, (0 missing)
  Item_Weight < 12.625 to the left, improve=0.0302976900,
(0 missing)
  Item_Visibility < 0.09764521 to the right, improve=0.0202096200,
(0 missing)
  as.factor(Outlet_Size_Small) splits as LR, improve=0.0001383401, (0 missing)
  as.factor(`Outlet_Location_Type_Tier 1`) splits as LR, improve=0.0001383401, (0 missing)
Surrogate splits:
  as.factor(Outlet_Size_Small) splits as RL, agree=0.829, adj=0.490, (0 split)
  as.factor(`Outlet_Location_Type_Tier 1`) splits as RL, agree=0.829, adj=0.490, (0 split)
  as.factor(`Outlet_Location_Type_Tier 3`) splits as LR, agree=0.829, adj=0.490, (0 split)
  Item_Visibility < 0.008760024 to the right, agree=0.669, adj=0.01
2, (0 split)
  Item_Weight < 4.9 to the right, agree=0.665, adj=0.00
1, (0 split)

```

```

Node number 21: 1856 observations,    complexity param=0.002451204
mean=43.96625, MSE=238.8889
left son=42 (927 obs) right son=43 (929 obs)
Primary splits:
  as.factor(`Outlet_Location_Type_Tier 1`) splits as LR, improve=0.015880820, (0 missing)
  as.factor(`Outlet_Location_Type_Tier 3`) splits as RL, improve=0.015880820, (0 missing)
  Item_Visibility < 0.1599065 to the right, improve=0.001985009,
(0 missing)
  Item_Weight < 16.925 to the left, improve=0.001082007,
(0 missing)
Surrogate splits:
  Item_Weight < 8.305 to the left, agree=0.513, adj=0.026, (0 split)
  Item_Visibility < 0.0418986 to the right, agree=0.512, adj=0.023, (0 split)

Node number 40: 1848 observations,    complexity param=0.03364318
mean=31.32921, MSE=334.7085
left son=80 (475 obs) right son=81 (1373 obs)
Primary splits:
  as.factor(Outlet_Size_Small) splits as LR, improve=0.20427800, (0 missing)
  as.factor(`Outlet_Location_Type_Tier 1`) splits as LR, improve=0.20427800, (0 missing)
  as.factor(`Outlet_Location_Type_Tier 3`) splits as RL, improve=0.20427800, (0 missing)
  Item_Weight < 12.55 to the right, improve=0.04010251,
(0 missing)
  Item_Visibility < 0.09722637 to the right, improve=0.02612810,
(0 missing)
Surrogate splits:
  Item_Weight < 21.3 to the right, agree=0.744, adj=0.002, (0 split)

Node number 41: 932 observations
mean=45.17021, MSE=258.6473

Node number 42: 927 observations
mean=42.0164, MSE=229.8928

Node number 43: 929 observations
mean=45.91191, MSE=240.2863

Node number 80: 475 observations
mean=17.27092, MSE=45.40423

Node number 81: 1373 observations,    complexity param=0.03364318
mean=36.19279, MSE=342.7676
left son=162 (914 obs) right son=163 (459 obs)
Primary splits:
  Item_Weight < 12.55 to the right, improve=0.12279120, (0 missing)
  Item_Visibility < 0.1056321 to the right, improve=0.03047616, (0 missing)
Surrogate splits:
  Item_Visibility < 0.01167985 to the right, agree=0.671, adj=0.015, (0 split)

Node number 162: 914 observations,    complexity param=0.03364318
mean=31.59534, MSE=336.4522
left son=324 (450 obs) right son=325 (464 obs)
Primary splits:
  Item_Weight < 12.625 to the left, improve=0.55950920, (0 missing)

```

```
Item_Visibility < 0.1054815 to the right, improve=0.03801279, (0 missing)
Surrogate splits:
Item_Visibility < 0.05747654 to the right, agree=0.606, adj=0.2, (0 split)
```

```
Node number 163: 459 observations
mean=45.34761, MSE=229.4434
```

```
Node number 324: 450 observations
mean=17.66319, MSE=57.44482
```

```
Node number 325: 464 observations
mean=45.10712, MSE=236.225
```

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```
main_predict2 <- predict(tree1, newdata = new_test_combi_encoded_dummies_drop, type = "vector")
sub_file <- data.frame(Item_Identifier = test$Item_Identifier, Outlet_Identifier = test$Outlet_I
dentifier, Item_Outlet_Sales = main_predict2)
write.csv(sub_file, 'Decision_tree_sales.csv')
```

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```
main_predict3 <- predict(tree2, newdata = new_test_combi_encoded_dummies_drop, type = "vector")
sub_file <- data.frame(Item_Identifier = test$Item_Identifier, Outlet_Identifier = test$Outlet_I
dentifier, Item_Outlet_Sales = main_predict3)
write.csv(sub_file, 'Decision_tree_sales_sqrt.csv')
```