npillana_Assignment 1

Loading Libraries

```
library(dplyr)

##

## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':

##

## filter, lag

## The following objects are masked from 'package:base':

##

## intersect, setdiff, setequal, union

library(ISLR)

library(ggplot2)
```

Data Preparation

##Loading DataSet

```
SafeBabies <- Carseats %>% select("Sales", "Price", "ShelveLoc")
summary(SafeBabies)
##
       Sales
                       Price
                                   ShelveLoc
## Min. : 0.000
                   Min. : 24.0
                                  Bad : 96
## 1st Qu.: 5.390
                   1st Qu.:100.0
                                  Good : 85
                   Median :117.0
## Median : 7.490
                                  Medium:219
## Mean : 7.496
                   Mean :115.8
## 3rd Qu.: 9.320
                   3rd Qu.:131.0
## Max. :16.270
                  Max. :191.0
```

Clean the data. Check for any missing vales

```
table(is.na(SafeBabies))
##
## FALSE
## 1200
sum(is.na(SafeBabies))
## [1] 0
```

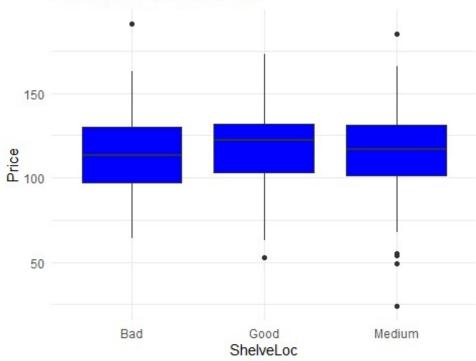
Divide data based on shelveLoc i.e, Good, Bad and Medium

```
Good <- filter(SafeBabies, ShelveLoc == "Good")
Bad <- filter(SafeBabies, ShelveLoc == "Bad")
Medium <- filter(SafeBabies, ShelveLoc == "Medium")</pre>
```

Visualizing the Price and Sales based on Shelve location

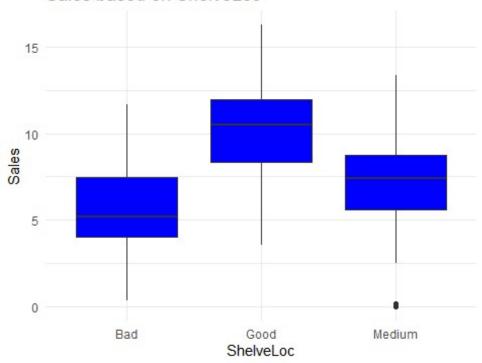
```
ggplot(data=SafeBabies) + aes(x = ShelveLoc, y = Price) + geom_boxplot(fill =
"blue") + labs(x = "ShelveLoc", title = "Price based on ShelveLoc") +
theme_minimal()
```

Price based on ShelveLoc



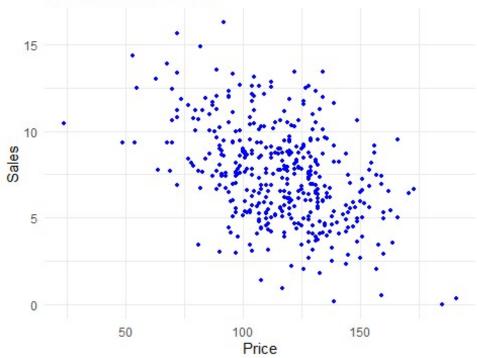
```
ggplot(data=SafeBabies) + aes(x = ShelveLoc, y = Sales) + geom_boxplot(fill =
"blue") + labs(x = "ShelveLoc", title = "Sales based on ShelveLoc") +
theme_minimal()
```

Sales based on ShelveLoc



ggplot(data=SafeBabies) + aes(x = Price, y = Sales) + geom_point(size = 1L,
colour = "blue") + labs(title = "Sales based on Price") + theme_minimal()

Sales based on Price



Based on above visuals the ShelveLoc with Good has the highest sales when compared to others and the price for Good ShelveLoc is slightly higher than Bad and Medium

Build a Linear Model

Lets build a linear model to predict the sales for all ShelveLoc

```
Good_Shelve <- lm(Sales ~ Price, data = Good)</pre>
Bad_Shelve <- lm(Sales ~ Price, data = Bad)</pre>
summary(Good Shelve)
##
## Call:
## lm(formula = Sales ~ Price, data = Good)
## Residuals:
##
      Min
              10 Median
                            30
                                  Max
## -3.721 -1.351 -0.098 1.483 4.353
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 17.968864
                                    18.187 < 2e-16 ***
                           0.988008
                           0.008199 -8.023 5.85e-12 ***
## Price
               -0.065785
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 1.888 on 83 degrees of freedom
## Multiple R-squared: 0.4368, Adjusted R-squared:
## F-statistic: 64.37 on 1 and 83 DF, p-value: 5.848e-12
summary(Bad_Shelve)
##
## Call:
## lm(formula = Sales ~ Price, data = Bad)
##
## Residuals:
      Min
                1Q Median
##
                                3Q
                                       Max
## -4.4622 -1.0617 -0.2014 1.2050 4.6412
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
                           0.990317 11.949 < 2e-16 ***
## (Intercept) 11.832984
## Price
               -0.055220
                           0.008486 -6.507 3.7e-09 ***
## ---
## Signif. codes:
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 1.967 on 94 degrees of freedom
## Multiple R-squared: 0.3105, Adjusted R-squared: 0.3032
## F-statistic: 42.34 on 1 and 94 DF, p-value: 3.702e-09
```

The adjusted R2 tells you the percentage of variation explained by only the independent variables that actually affect the dependent variable. Adjusted R-squared:0.43 for Good ShelveLoc and Adjusted R-squared:0.3032 for Bad ShelveLoc, which explains variability in the car price for Car Seats.

Formulating the Problem

Total Profit= Sales * Profit_per_Unit Profit_per_Unit= price - cost Sales=b1*price+b0 Total* Profit= (price - cost) (b1price+b0) =b1price^2+b0price-costb1price-costb0 =b1price^2+(b0-costb1)price-costb0

Now we have to find the optimal value, set the derivative to Zero Total Profit derivative = 2b1price + b0-b1cost

Then Price is; Price = (-b0)-(b1*cost))/2b1

Solving the Problem

```
#Lets calculate optimal price for car seat for both Good and Bad Shelve at
cost $55
OptimalPrice <- function(prod_cost, b_0, b_1) {return(((-1 * b_0) + (b_1 *
prod cost)) /(2 * b 1))}
Good OptimalPrice<- OptimalPrice(55, Good Shelve$coefficients[1],</pre>
Good Shelve$coefficients[2])
Good OptimalPrice
## (Intercept)
##
      164.0731
paste("optimal price for car seats with good shelve",Good_OptimalPrice)
## [1] "optimal price for car seats with good shelve 164.07312564386"
Bad OptimalPrice<- OptimalPrice(55, Bad Shelve$coefficients[1],</pre>
Bad_Shelve$coefficients[2])
Bad OptimalPrice
## (Intercept)
      134,6435
##
paste("optimal price for car seats with Bad shelve", Bad OptimalPrice)
## [1] "optimal price for car seats with Bad shelve 134.643464696399"
```

```
Calculate the optimal prive when the cost varies from $40 to $80 with $5 increments

Varied_Cost <- seq(40,85, by=5)

for( i in Varied_Cost){
```

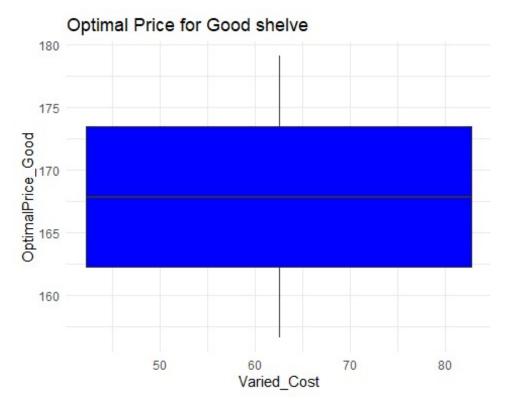
```
OptimalPrice_Good <- OptimalPrice(Varied_Cost,
Good_Shelve$coefficients[1],Good_Shelve$coefficients[2])
   OptimalPrice_Bad <- OptimalPrice(Varied_Cost,
Bad_Shelve$coefficients[1],Bad_Shelve$coefficients[2])
}</pre>
```

Now, we will combine production costs and optimal prices for good and bad shelve location

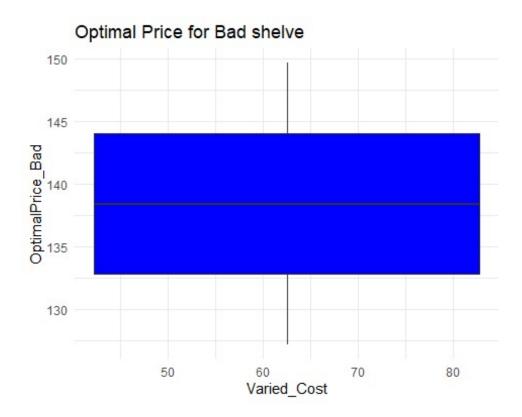
```
Combine cost <- data.frame(cbind(Varied Cost,OptimalPrice Good,</pre>
OptimalPrice Bad))
Combine cost
##
      Varied Cost OptimalPrice Good OptimalPrice Bad
## 1
                40
                             156.5731
                                               127.1435
## 2
                45
                             159.0731
                                               129.6435
## 3
                50
                             161.5731
                                               132.1435
## 4
                55
                             164.0731
                                               134.6435
## 5
                60
                             166.5731
                                               137.1435
                65
                                               139.6435
## 6
                             169.0731
## 7
                70
                             171.5731
                                               142.1435
                75
## 8
                             174.0731
                                               144.6435
## 9
                80
                             176.5731
                                               147.1435
                                               149.6435
## 10
                85
                             179.0731
```

Visualizing the optimal cost with varying with Production costs from \$40 to \$85 for Good Shelve and Bad Shelve

```
ggplot(data=Combine_cost) + aes(x = Varied_Cost, y = OptimalPrice_Good) +
geom_boxplot(fill = "blue") + labs(x = "Varied_Cost", title = "Optimal Price
for Good shelve") + theme_minimal()
## Warning: Continuous x aesthetic -- did you forget aes(group=...)?
```



```
ggplot(data=Combine_cost) + aes(x = Varied_Cost, y = OptimalPrice_Bad) +
geom_boxplot(fill = "blue") + labs(x = "Varied_Cost", title = "Optimal Price
for Bad shelve") + theme_minimal()
## Warning: Continuous x aesthetic -- did you forget aes(group=...)?
```



The above Visuals show that the Optimal Price for the Car seats at Good Shelve is higher than the Optimal Price for the Car seats at Bad Shelve.