

Assignment 1

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```
library(ISLR)
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(ggplot2)
library(tidyverse)

## -- Attaching packages -----
## ----- tidyverse 1.3.0 --

## v tibble  2.1.3      v purrr   0.3.3
## v tidyr   1.0.0      v stringr 1.4.0
## v readr   1.3.1      v forcats 0.4.0

## -- Conflicts -----
## ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()

SafeBabies <- Carseats %>% select("Sales", "Price", "ShelveLoc")
head(SafeBabies)

##   Sales Price ShelveLoc
## 1  9.50  120        Bad
## 2 11.22   83        Good
## 3 10.06   80    Medium
## 4  7.40   97    Medium
## 5  4.15  128        Bad
## 6 10.81   72        Bad

good<-SafeBabies%>%filter(ShelveLoc=="Good")
head(good)
```

```
## Sales Price ShelfLoc
## 1 11.22 83 Good
## 2 11.85 120 Good
## 3 11.96 94 Good
## 4 10.96 86 Good
## 5 11.17 118 Good
## 6 7.58 110 Good
```

```
medium<-SafeBabies%>%filter(ShelveLoc=="Medium")
head(medium)
```

```
## Sales Price ShelfLoc
## 1 10.06 80 Medium
## 2 7.40 97 Medium
## 3 6.63 108 Medium
## 4 6.54 124 Medium
## 5 4.69 124 Medium
## 6 3.98 136 Medium
```

```
bad<-SafeBabies%>%filter(ShelveLoc=="Bad")
head(bad)
```

```
## Sales Price ShelfLoc
## 1 9.50 120 Bad
## 2 4.15 128 Bad
## 3 10.81 72 Bad
## 4 9.01 100 Bad
## 5 10.14 113 Bad
## 6 2.99 97 Bad
```

model building

We are building linear model instead of building the generalized linear model because it makes no sense to build a generalized linear model on a single variable

```
modelgood<-lm(Sales~Price,data=good)
summary(modelgood)
```

```
##
## Call:
## lm(formula = Sales ~ Price, data = good)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.721 -1.351 -0.098  1.483  4.353
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 17.968864   0.988008  18.187  < 2e-16 ***
## Price       -0.065785   0.008199  -8.023 5.85e-12 ***
## ---
```

```
## 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:  0.43
## F-statistic: 64.37 on 1 and 83 DF,  p-value: 5.848e-12

modelbad<-lm(Sales~Price,data=bad)
summary(modelbad)

##
## 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|)
## (Intercept) 11.832984   0.990317  11.949  < 2e-16 ***
## 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
```

optimal cost for shelf location good

```
Productioncost<-55
paste("The optimal cost for shelf location good", ((-
modelgood$coefficients[[2]] *Productioncost) +
(modelgood$coefficients[[1]]))/(-2 * modelgood$coefficients[[2]]))

## [1] "The optimal cost for shelf location good 164.07312564386"
```

optimal cost for shelf location bad

```
paste("The optimal cost for shelf location bad", ((-
modelbad$coefficients[[2]] *Productioncost) +
(modelbad$coefficients[[1]]))/(-2 * modelbad$coefficients[[2]]))

## [1] "The optimal cost for shelf location bad 134.643464696399"
```

plotting optimal cost for varying production costs

```
productioncost1<-data.frame(prodncost=seq(40,85,by=1),optimalprice=rep(0,46))
for (i in 1:nrow(productioncost1)) {
  productioncost1$optimalprice[i]<- ((-modelgood$coefficients[[2]]
*productioncost1$prodncost[i]) + (modelgood$coefficients[[1]]))/(-2 *
modelgood$coefficients[[2]])
}
```

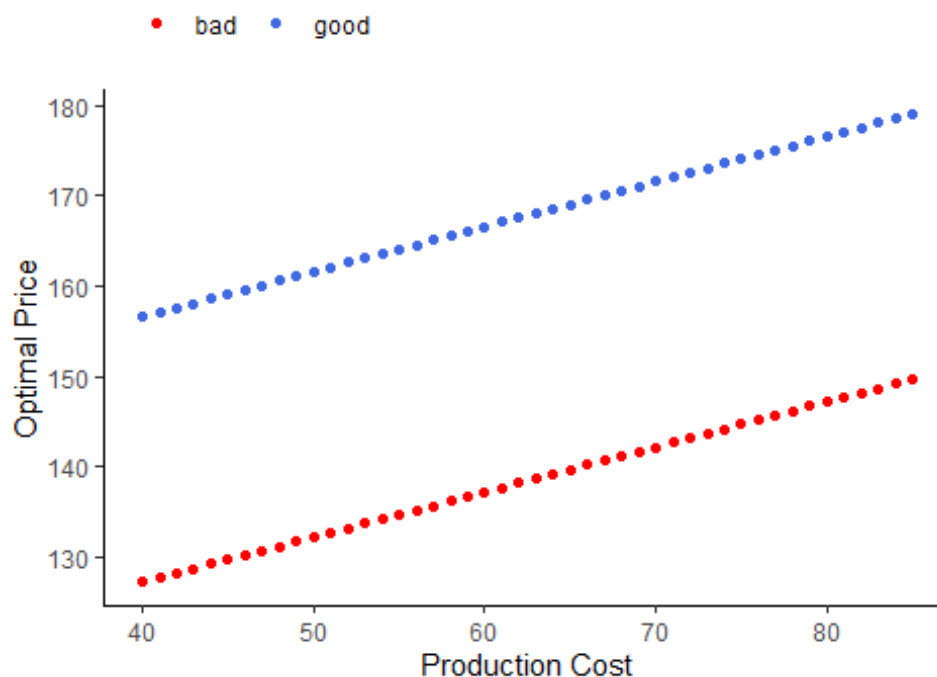
```

productioncost2 <- c(40:85)
rr <- 0
for (i in 1:length(productioncost2)) {
  rr[i] <- ((-modelbad$coefficients[[2]] * productioncost2[i]) +
(modelbad$coefficients[[1]]))/(-2 * modelbad$coefficients[[2]])
}
productioncost2<-data.frame(prodncost=seq(40,85,by=1),optimalprice=rr)
data<-cbind(productioncost1,productioncost2$optimalprice)
names(data)<-c("productioncost","good","bad")
data1<- data %>% gather("category","optimalprice",-productioncost)

ggplot(data1,aes(x=productioncost,y=optimalprice,color=category))+
  geom_point()+theme_classic()+ labs(x="Production Cost",y="Optimal
Price",color="Shelve Location") +
  ggtitle("Optimal Prices for Shelve Location")+
  theme(legend.position = "top",legend.justification = "left",legend.title =
element_blank()) + scale_color_manual(values=c("red1","royalblue"))

```

Optimal Prices for Shelve Location



We can observe from the above plot that if the Shelve Location is good the optimal price for the product is greater than the products placed at shleve location bad, That makes sense!!!