MA331- Final Exam

Project Continuation:

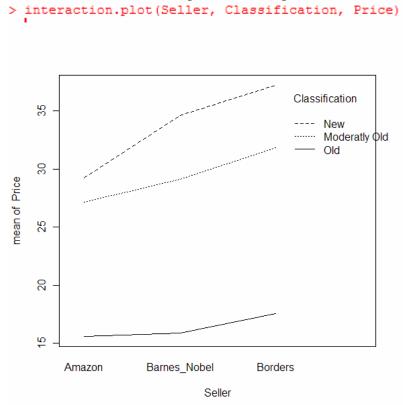
Before running the two-way ANOVA tests we check for interaction between the variables for the two cases, one including the age of the books and the other including the Price classification (i.e. expensive)

• We first check to make sure that both variables are factors or in other words categorical.

```
> getwd()
[1] "D:/Documents and Settings/Class2009/My Documents"
> setwd("D:/Documents and Settings/Class2009/My Documents/Fall08/MA331")
> Data <- read.csv("Project2.csv", header = TRUE)
> attach(Data)
> is.factor(Classification)
[1] TRUE
> is.factor(Seller:)
[1] TRUE
```

 Next we take a look at the Interaction Plot. For Seller/Vendor and Age Classification

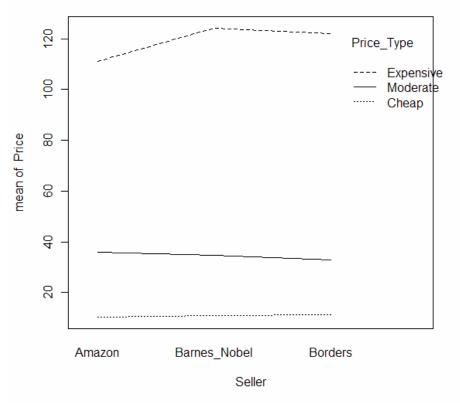
Note: in the below plot formula, the first two arguments are the cathegorical variables and the last is the quantitative response



From the graph, we can gather that,

- There is little to no interaction between the Seller and the Age Classification of the books.
- We see that the Age appears to have some significant influence on the mean of Price
- As was found in the earlier studies, the average price of Borders books is a bit higher than that of Barnes and Nobel's, which is in tern a bit higher than Amazon.
- Next we take a look at the Interaction Plot. For Seller/Vendor and Price Type (i.e. expensive)

> interaction.plot(Seller, Price_Type, Price)



From the graph, we can gather that,

- There is little to no interaction between the Seller and the Price type of the books.
- We see that the mean for the price of Expensive Amazon books seems significantly lower than that of both Barnes and Nobel and Borders.

The analysis of the graphs was based off of the following observations from the lecture.

Γ	Neither factor is significant	Neither factor is significant	Only 1 factor is significant	Both factors are significant
	No interaction	Interaction effect is significant	No interaction	With or without significant interaction
Dependent var.		\succ		

Levels of factor A Levels of factor A

Verification of the previous observations

The below summary clearly indicates that there is not interaction between seller and age of books, with effect on the price. This means that for that with regards to these two variables a one-way analysis is sufficient.

Analysis of Variance Table:

```
> anova(lm(Price~Seller+Classification+Seller:Classification))
  Analysis of Variance Table
   Response: Price
                        Df Sum Sg Mean Sg F value Pr(>F)
                        2 664 332 0.2164 0.8057
   Seller
   Classification 2 6635 3317 2.1625 0.1188
  Seller:Classification 4 161 40 0.0262 0.9987
  Residuals 141 216302 1534
> summary(lm(Price~Seller+Classification+Seller:Classification))
Call:
lm(formula = Price ~ Seller + Classification + Seller:Classification)
Residuals:
   Min
          1Q Median
                       3Q
                               Max
-30.197 -21.201 -11.828 -1.738 172.846
Coefficients:
                                Estimate Std. Error t value Pr(>|t|)
                                 27.154 8.547 3.177 0.00183 **
(Intercept)
SellerBarnes Nobel
                                          12.087 0.163 0.87086
                                  1.969
SellerBorders
                                  4.669
                                          12.087 0.386 0.69990
ClassificationNew
                                  2.037
                                          12.778 0.159 0.87358
ClassificationOld
                                -11.554
                                          14.174 -0.815 0.41633
SellerBarnes_Nobel:ClassificationNew 3.460 18.071 0.191 0.84843
SellerBorders:ClassificationNew
                                 3.327 18.071 0.184 0.85418
SellerBarnes_Nobel:ClassificationOld -1.704 20.044 -0.085 0.93236
SellerBorders:ClassificationOld -2.708 20.044 -0.135 0.89274
Signif. codes: 0 `***' 0.001 `**' 0.01 `*' 0.05 `.' 0.1 ` ' 1
Residual standard error: 39.17 on 141 degrees of freedom
Multiple R-squared: 0.03334, Adjusted R-squared: -0.02151
```

F-statistic: 0.6078 on 8 and 141 DF, p-value: 0.7702

According to the below summary there seems to be interaction between seller and Price_Type, with effect on the price. This is indicated by the strong P-value in the "Seller:Price_Type" row of the ANOVA results. However, by observing the summary of the lm we find that the "Price_TypeExpensive" and Price_TypeModerate" rows are the two rows having a strong effect. This is expected as a Price categorical value for price should follow the general trend of the mean prices for the books.

Analysis of Variance Table:

```
> anova(lm(Price~Seller+Price Type+Seller:Price Type))
     Analysis of Variance Table
      Response: Price
                                        Df Sum Sq Mean Sq F value Pr(>F)
     Seller
                                        2 664 332 1.4885 0.2292
     Price_Type 2 191097 95548 428.4367 <2e-16 ***
     Seller:Price_Type 4 556 139 0.6229 0.6469
Residuals 141 31445 223
      Signif. codes: 0 `***' 0.001 `**' 0.01 `*' 0.05 `.' 0.1 ` ' 1
> summary(lm(Price~Seller+Price Type+Seller:Price Type))
Call:
lm(formula = Price ~ Seller + Price Type + Seller:Price Type)
Residuals:
     Min 1Q Median 3Q
                                                          Max
-37.177 -3.891 -2.486 3.121 89.003
Coefficients:
                                                                  Estimate Std. Error t value Pr(>|t|)
                                                                  10.4762 2.3913 4.381 2.29e-05 ***
0.4049 3.4272 0.118 0.906117
 (Intercept)
SellerBarnes_Nobel

      SellerBarnes_Nobel
      0.4049
      3.4272
      0.118
      0.906117

      SellerBorders
      0.7601
      3.4771
      0.219
      0.827269

      Price_TypeExpensive
      100.5205
      6.5489
      15.349
      < 2e-16</td>
      ***

      Price_TypeModerate
      25.3518
      7.0938
      3.574
      0.000482
      ***

      SellerBarnes_Nobel:Price_TypeExpensive
      12.7301
      9.2782
      1.372
      0.172228

      SellerBorders:Price_TypeExpensive
      10.4203
      9.0066
      1.157
      0.249242

      SellerBarnes_Nobel:Price_TypeModerate
      -1.5044
      9.3919
      -0.160
      0.872972

      SellerBorders:Price_TypeModerate
      -3.7531
      9.1962
      -0.408
      0.683807

Signif. codes: 0 `***' 0.001 `**' 0.01 `*' 0.05 `.' 0.1 ` ' 1
Residual standard error: 14.93 on 141 degrees of freedom
Multiple R-squared: 0.8595, Adjusted R-squared: 0.8515
```

F-statistic: 107.8 on 8 and 141 DF, p-value: < 2.2e-16

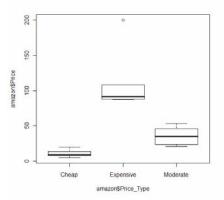
Do to the findings in the summary for he interaction between vendor and price type we will look at the individual graphs of the price~price type for each vendor. We specifically need to pay attention to the behavior for Expensive and moderate books.

```
> plot(bn$Price~bn$Price_Type)
> plot(amazon$Price~amazon$Price Type)
```

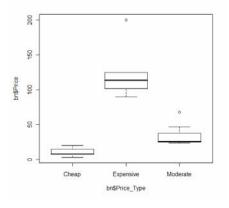
```
> plot(amazon$File="amazon$File="type")
> plot(borders$Price-borders$Price Type)
```

, bioc(porders@filde~porders@filde

Amazon: Price~Price_Type





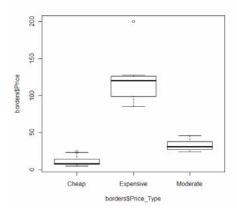


From these Graphs & the Graphs that follow:

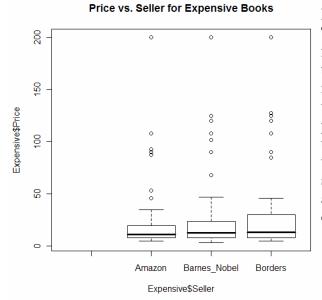
It is evident that the behavior of the three vendors varies in the case of books which are classified as either Expensive or Mederate. A buyer can expect to be much more likely to pay more for Expensive books if they are buying from Borders. A middle ground lies in Barnes & Nobel and a much lower range with Amazon.

In contrast to my earlier findings without the use of the two-way ANOVA, it seems from this data that my large amount of "Cheap" books was masking the variance that could be found in the smaller data sets of Expensive and Moderate books.

Borders: Price~Price_Type



ExpensiveBooks:



ModerateBooks:

200 0 0 0 150 Moderate\$Price 8 8 0 00 0 100 8 0 00 22 0 Barnes Nobel Borders Amazon Moderate\$Seller

Price vs. Seller for Moderate Books

Reflection:

This project has been extremely useful; I feel that I have learned a great deal about the overall process of statistics. It was interesting to pick a problem and just go with it. I was happy to have chosen a project in which I could gain an understanding of ANOVA as apposed to working on a problem where I would simply be using the same regression analysis, which was the case for many other groups.