

# STA 4320 CHAP 3.3.2

Prof. He Jiang

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## Sec 3.3.2

Advertising dataset

```
fpath = getwd()
Advertising = read.csv(paste0(fpath, "/Advertising.csv"))
```

Credit and Auto dataset

```
require(ISLR2)
```

```
## Loading required package: ISLR2
```

## Interaction

Advertising interaction between TV and radio

```
reg = lm(sales ~ TV * radio, data = Advertising)
# equivalent to
# reg = lm(sales ~ TV + radio + TV:radio, data = Advertising)
```

```
summary(reg)
```

```
##
## Call:
## lm(formula = sales ~ TV * radio, data = Advertising)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -6.3366 -0.4028  0.1831  0.5948  1.5246
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  6.750e+00  2.479e-01  27.233  <2e-16 ***
## TV           1.910e-02  1.504e-03  12.699  <2e-16 ***
## radio        2.886e-02  8.905e-03   3.241   0.0014 **
## TV:radio     1.086e-03  5.242e-05  20.727  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9435 on 196 degrees of freedom
## Multiple R-squared:  0.9678, Adjusted R-squared:  0.9673
## F-statistic: 1963 on 3 and 196 DF, p-value: < 2.2e-16
```

Compare to model with only TV and radio separately

```
reg_0 = lm(sales ~ TV + radio, data = Advertising)
summary(reg_0)$r.sq
```

```
## [1] 0.8971943
```

```
# note in the increase in R2 for the interaction model
```

Regression of Balance on Income and Student without interaction

$Y = \text{beta\_0} + \text{beta\_1} * \text{income} + \text{beta\_2} * \text{student} + \text{epsilon}$

```
head(Credit)
```

```
##      Income Limit Rating Cards Age Education Own Student Married Region Balance
## 1  14.891  3606    283    2  34         11 No      No      Yes  South    333
## 2 106.025  6645    483    3  82         15 Yes     Yes     Yes   West    903
## 3 104.593  7075    514    4  71         11 No      No      No    West    580
## 4 148.924  9504    681    3  36         11 Yes     No      No    West    964
## 5  55.882  4897    357    2  68         16 No      No      Yes   South    331
## 6  80.180  8047    569    4  77         10 No      No      No    South   1151
```

```
reg = lm(Balance ~ Income + Student, data = Credit)
summary(reg)
```

```
##
```

```
## Call:
```

```
## lm(formula = Balance ~ Income + Student, data = Credit)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
## -762.37 -331.38  -45.04   323.60   818.28
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  211.1430    32.4572   6.505 2.34e-10 ***
## Income         5.9843     0.5566  10.751 < 2e-16 ***
## StudentYes    382.6705    65.3108   5.859 9.78e-09 ***
```

```
## ---
```

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

```
##
```

```
## Residual standard error: 391.8 on 397 degrees of freedom
```

```
## Multiple R-squared:  0.2775, Adjusted R-squared:  0.2738
```

```
## F-statistic: 76.22 on 2 and 397 DF,  p-value: < 2.2e-16
```

Plot of the two lines with same slope

```
beta_0_hat = as.numeric( reg$coefficients[1] )
beta_1_hat = as.numeric( reg$coefficients[2] )
beta_2_hat = as.numeric( reg$coefficients[3] )
```

```
plot(Credit$Income, Credit$Balance,
     pch = 16,
     xlab = "Income",
     ylab = "Balance")
```

```
# student line
```

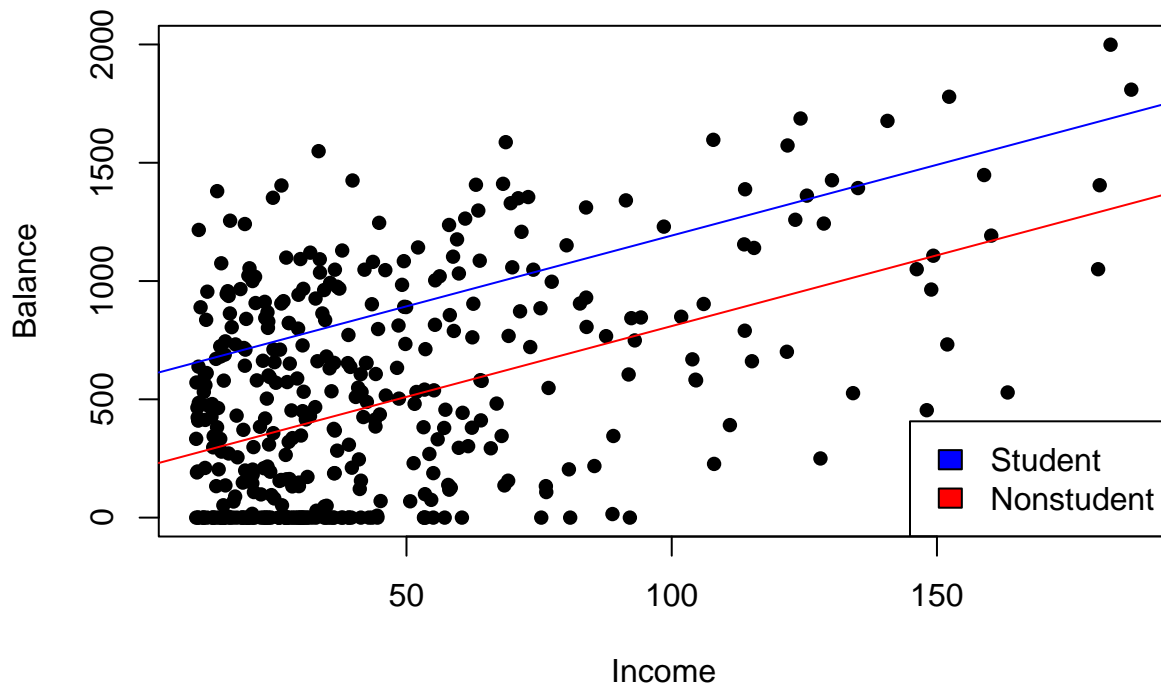
```
abline(a = beta_0_hat + beta_2_hat,
```

```

    b = beta_1_hat,
    col = "blue")
# nonstudent line
abline(a = beta_0_hat,
       b = beta_1_hat,
       col = "red")

legend("bottomright",
      legend = c("Student", "Nonstudent"),
      fill = c("blue", "red"))

```



Regression of Balance on Income and Student with interaction

$$Y = \beta_0 + \beta_1 * \text{income} + \beta_2 * \text{student} + \beta_3 * \text{income} * \text{student} + \text{epsilon}$$

```

reg = lm(Balance ~ Income + Student + Income:Student, data = Credit)
summary(reg)

```

```

##
## Call:
## lm(formula = Balance ~ Income + Student + Income:Student, data = Credit)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -773.39 -325.70  -41.13   321.65   814.04
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    200.6232     33.6984   5.953 5.79e-09 ***
## Income           6.2182      0.5921  10.502 < 2e-16 ***
## StudentYes     476.6758    104.3512   4.568 6.59e-06 ***
## Income:StudentYes -1.9992     1.7313  -1.155  0.249
## ---

```

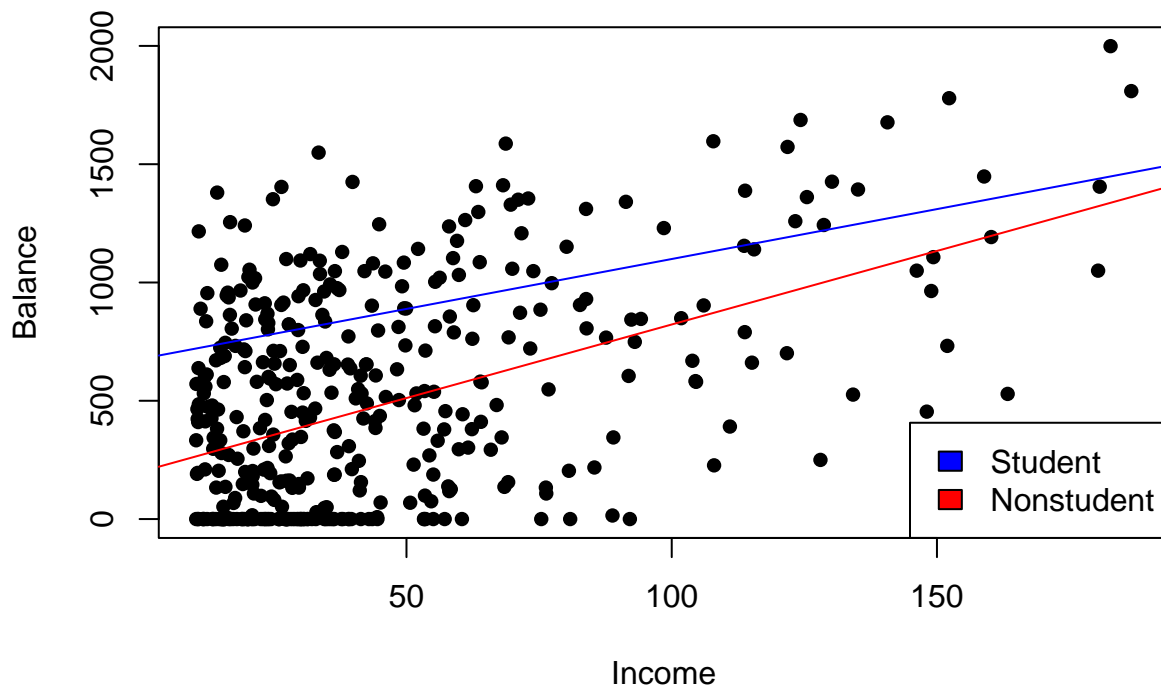
```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 391.6 on 396 degrees of freedom
## Multiple R-squared:  0.2799, Adjusted R-squared:  0.2744
## F-statistic: 51.3 on 3 and 396 DF,  p-value: < 2.2e-16
```

Plot of the two lines with different slope

```
beta_0_hat = as.numeric( reg$coefficients[1] )
beta_1_hat = as.numeric( reg$coefficients[2] )
beta_2_hat = as.numeric( reg$coefficients[3] )
beta_3_hat = as.numeric( reg$coefficients[4] )

plot(Credit$Income, Credit$Balance,
     pch = 16,
     xlab = "Income",
     ylab = "Balance")
# student line
abline(a = beta_0_hat + beta_2_hat,
      b = beta_1_hat + beta_3_hat,
      col = "blue")
# nonstudent line
abline(a = beta_0_hat,
      b = beta_1_hat,
      col = "red")

legend("bottomright",
      legend = c("Student", "Nonstudent"),
      fill = c("blue", "red"))
```



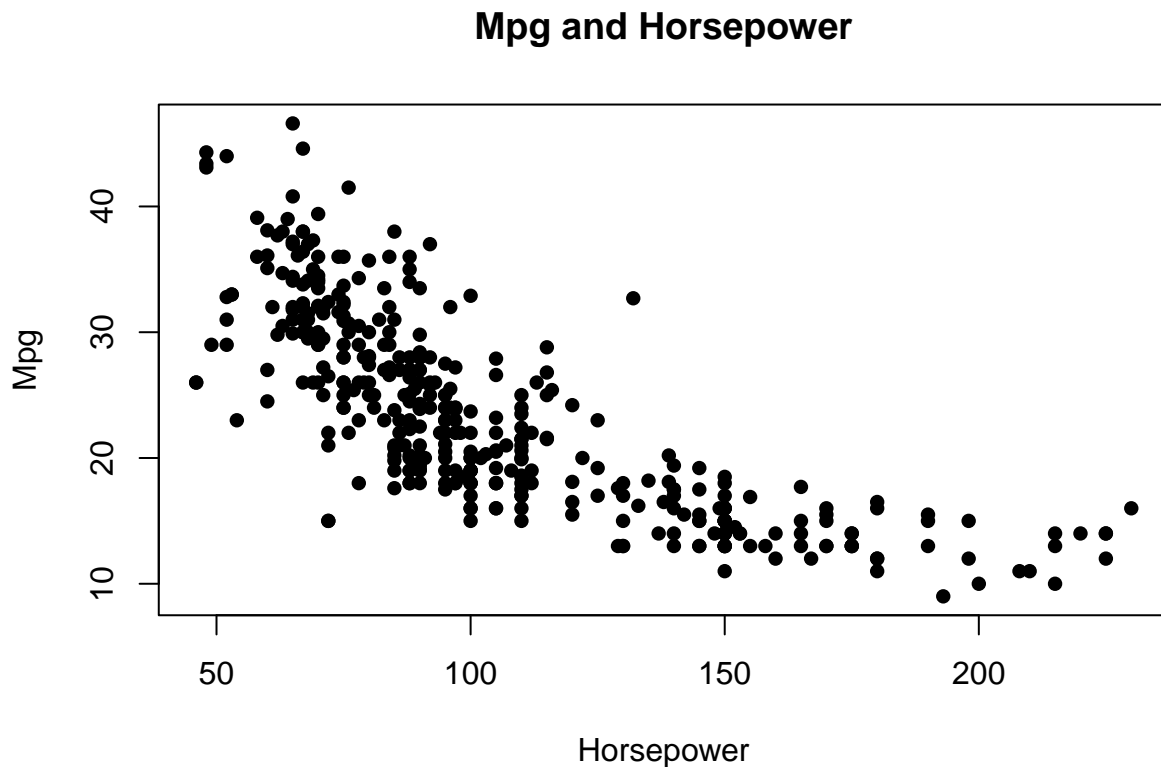
## Polynomial

Auto: mpg vs horsepower

Scatterplot

```
y = Auto$mpg
x = Auto$horsepower
```

```
plot(x, y,
     main = "Mpg and Horsepower",
     pch = 16,
     xlab = "Horsepower",
     ylab = "Mpg")
```



Polynomial of degree 2

```
reg_2 = lm(mpg ~ horsepower + I(horsepower^2), data = Auto)
# alternatively here reg_2 = lm(y ~ x + I(x^2))
summary(reg_2)
```

```
##
## Call:
## lm(formula = mpg ~ horsepower + I(horsepower^2), data = Auto)
##
## Residuals:
```

	Min	1Q	Median	3Q	Max
##	-14.7135	-2.5943	-0.0859	2.2868	15.8961

```
##
## Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t )
##				

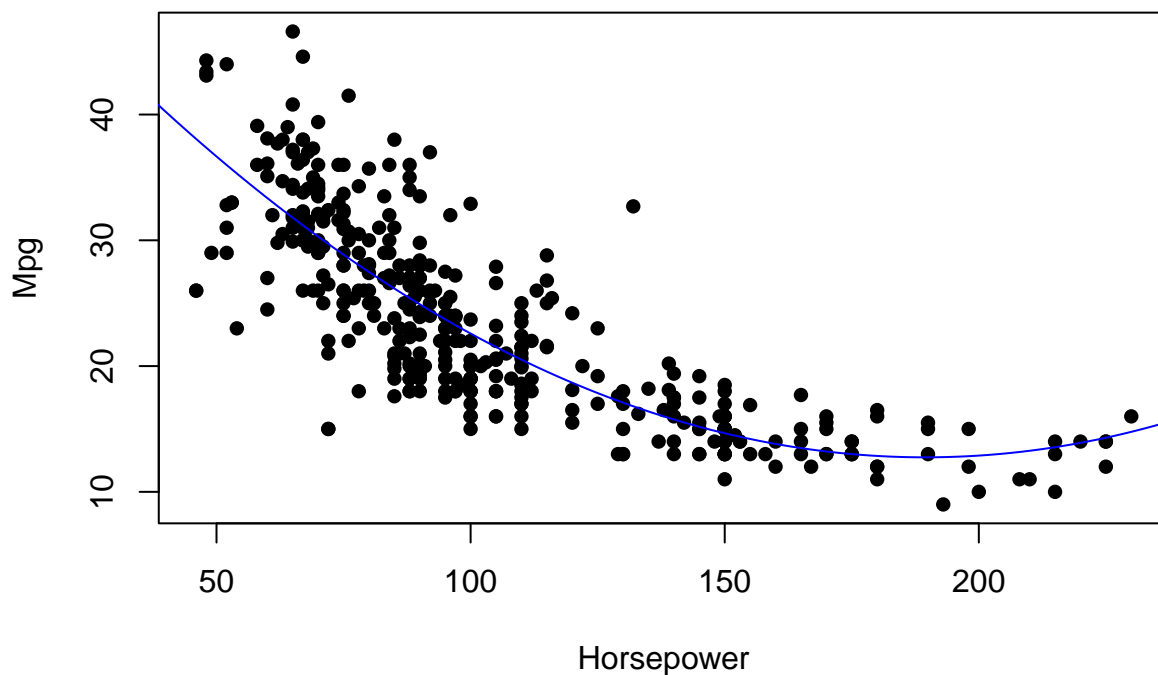
```
## (Intercept)      56.9000997  1.8004268   31.60   <2e-16 ***
## horsepower      -0.4661896  0.0311246  -14.98   <2e-16 ***
## I(horsepower^2)  0.0012305  0.0001221   10.08   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.374 on 389 degrees of freedom
## Multiple R-squared:  0.6876, Adjusted R-squared:  0.686
## F-statistic:   428 on 2 and 389 DF,  p-value: < 2.2e-16
```

Plot

```
plot(x, y,
     main = "Mpg and Horsepower",
     pch = 16,
     xlab = "Horsepower",
     ylab = "Mpg")

# plotting here requires the
x_poss = seq(min(x) - 10, max(x) + 10, length.out = 1000)
y_poss = predict(reg_2, data.frame(horsepower = x_poss))
lines(x_poss, y_poss, col = "blue")
```

## Mpg and Horsepower



An alternative degree 5 polynomial

```
reg_5 = lm(mpg ~ poly(horsepower, 5), data = Auto)
summary(reg_5)

##
## Call:
## lm(formula = mpg ~ poly(horsepower, 5), data = Auto)
##
```

```
## Residuals:
##      Min       1Q   Median       3Q      Max
## -15.4326  -2.5285  -0.2925   2.1750  15.9730
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      23.4459     0.2185  107.308 < 2e-16 ***
## poly(horsepower, 5)1 -120.1377     4.3259  -27.772 < 2e-16 ***
## poly(horsepower, 5)2  44.0895     4.3259   10.192 < 2e-16 ***
## poly(horsepower, 5)3  -3.9488     4.3259   -0.913  0.36190
## poly(horsepower, 5)4  -5.1878     4.3259   -1.199  0.23117
## poly(horsepower, 5)5   13.2722     4.3259    3.068  0.00231 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.326 on 386 degrees of freedom
## Multiple R-squared:  0.6967, Adjusted R-squared:  0.6928
## F-statistic: 177.4 on 5 and 386 DF,  p-value: < 2.2e-16
```

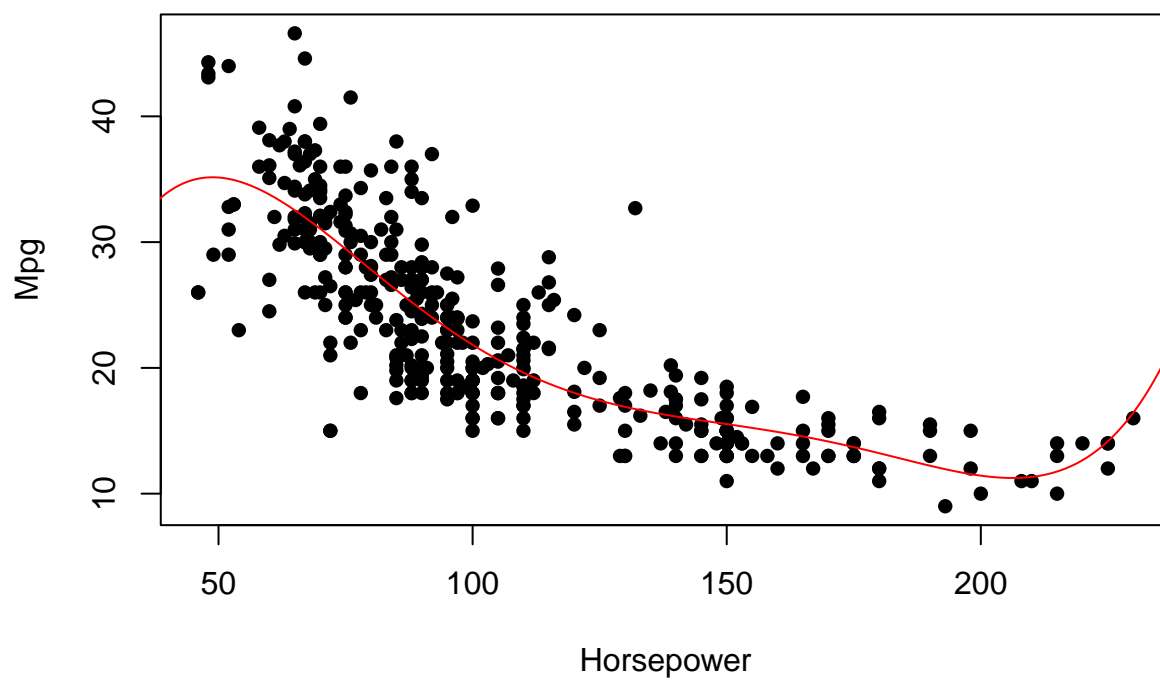
Plot

```
plot(x, y,
     main = "Mpg and Horsepower",
     pch = 16,
     xlab = "Horsepower",
     ylab = "Mpg")

# plotting here requires the
x_poss = seq(min(x) - 10, max(x) + 10, length.out = 1000)

y_poss = predict(reg_5, data.frame(horsepower = x_poss))
lines(x_poss, y_poss, col = "red")
```

## Mpg and Horsepower



```
# could also add the degree 2 curve in the same picture  
# lines(x_poss, predict(reg_2, data.frame(horsepower = x_poss)), col = "blue")
```