Regression Analytics

Elmy Luka

2022-11-13

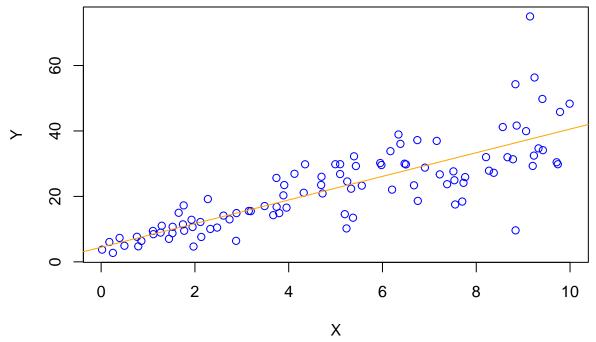
```
library('dplyr')
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
      filter, lag
## The following objects are masked from 'package:base':
      intersect, setdiff, setequal, union
##
library('tidyverse')
## -- Attaching packages ------ tidyverse 1.3.2 --
## v ggplot2 3.3.6
                     v purrr
                             0.3.4
## v tibble 3.1.8 v stringr 1.4.1
## v tidyr
          1.2.1
                   v forcats 0.5.2
## v readr
          2.1.2
                                           ----- tidyverse_conflicts() --
## -- Conflicts -----
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                   masks stats::lag()
library('tinytex')
```

#1. Run the following code in R-studio to create two variablesX and Y. #set.seed(2017) #X=runif(100) 10 #Y=X4+3.45 #Y=rnorm(100) 0.29 Y+Y

```
set.seed(2017)
X=runif(100)*10
Y=X*4+3.45
Y=rnorm(100)*0.29*Y+Y
```

#1a) Plot Y against X. Include a screenshot of the plot in your submission. Using the File menu you can save the graph as a picture on your computer. Based on the plot do you think we can fit a linear model to explain Y based on X?

```
plot(Y~X,xlab='X',ylab='Y',col='blue')
abline(lsfit(X, Y),col = "orange")
```



#From the plot it can be concluded that there is a correlation between the variables "x" and "y". Therefore linear model is a good fit as we can see a positive correlation between the attributes.

#1b) Construct a simple linear model of Y based on X. Write the equation that explains Y based on X. What is the accuracy of this model?

#The linear model of Y based on X is given by the equation Y = 4.4655 + 3.6108 * X #The accuracy of the model is 0.6517 or 65%. Additionally, this shows that X can account for 65.17 percent of the variation in Y.

```
linear_model <- lm(Y ~ X)
summary(linear_model)</pre>
```

```
##
## Call:
##
  lm(formula = Y \sim X)
##
## Residuals:
##
       Min
                                3Q
                                       Max
                1Q
                    Median
  -26.755 -3.846
                    -0.387
                                    37.503
##
                             4.318
##
##
   Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
##
                 4.4655
                            1.5537
                                     2.874
                                            0.00497 **
   (Intercept)
## X
                 3.6108
                            0.2666
                                    13.542
                                            < 2e-16 ***
##
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 7.756 on 98 degrees of freedom
## Multiple R-squared: 0.6517, Adjusted R-squared: 0.6482
## F-statistic: 183.4 on 1 and 98 DF, p-value: < 2.2e-16
```

#1c) How the Coefficient of Determination, R2, of the model above is related to the correlation coefficient of X and Y?

```
cor(X,Y)^2
```

[1] 0.6517187

#The coefficient of determination is equal to the square of the correlation coefficient. Therefore both the values of the coefficient of determination (R2) and the correlation coefficient of Y and X would be same.

#2. We will use the 'mtcars' dataset for this question. The dataset is already included in your R distribution. The dataset shows some of the characteristics of different cars. The following shows few samples (i.e. the first 6 rows) of the dataset. The description of the dataset can be found here.

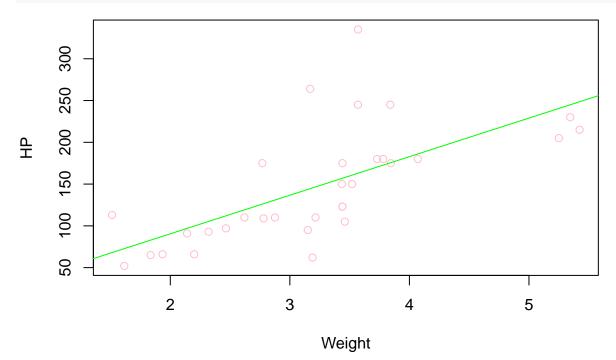
head(mtcars)

```
##
                                                      qsec vs am gear carb
                       mpg cyl disp
                                      hp drat
## Mazda RX4
                                     110 3.90 2.620
                                                     16.46
                                 160 110 3.90 2.875 17.02
                                                                     4
                                                                           4
## Mazda RX4 Wag
                      21.0
                              6
## Datsun 710
                      22.8
                                      93 3.85 2.320 18.61
                      21.4
                                                                     3
## Hornet 4 Drive
                                 258
                                     110 3.08 3.215 19.44
                              6
                                                                           1
                      18.7
                                     175 3.15 3.440 17.02
                                                                     3
                                                                           2
## Hornet Sportabout
                             8
                                                                     3
## Valiant
                      18.1
                                     105 2.76 3.460 20.22
                                                                           1
```

#2a) James wants to buy a car. He and his friend, Chris, have different opinions about the Horse Power (hp) of cars. James think the weight of a car (wt) can be used to estimate the Horse Power of the car while Chris thinks the fuel consumption expressed in Mile Per Gallon (mpg), is a better estimator of the (hp). Who do you think is right? Construct simple linear models using mtcars data to answer the question.

#Model from James estimation.

```
plot(mtcars$hp~mtcars$wt,xlab='Weight',ylab='HP',col='pink')
abline(lsfit(mtcars$wt,mtcars$hp),col = "green")
```

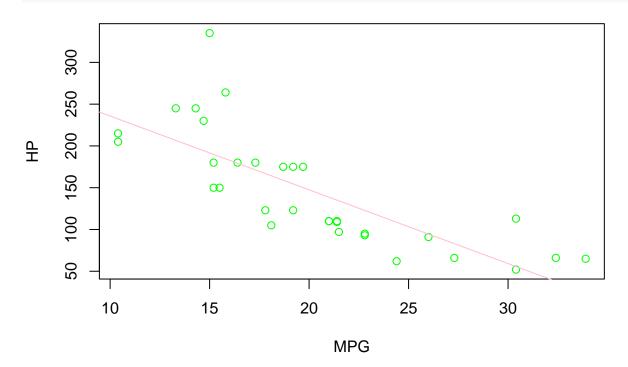


```
james_model<-lm(formula =hp~wt, data = mtcars )</pre>
summary(james_model)
##
## Call:
## lm(formula = hp ~ wt, data = mtcars)
##
## Residuals:
##
                1Q Median
       Min
                                 3Q
                                        Max
   -83.430 -33.596 -13.587
                             7.913 172.030
##
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
                 -1.821
                            32.325
                                    -0.056
                                               0.955
## (Intercept)
                 46.160
                                      4.796 4.15e-05 ***
## wt
                             9.625
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
\#\# Residual standard error: 52.44 on 30 degrees of freedom
## Multiple R-squared: 0.4339, Adjusted R-squared: 0.4151
                   23 on 1 and 30 DF, p-value: 4.146e-05
## F-statistic:
```

#This model has an accuracy of 0.4339

#Model from Chris estimation

```
plot(mtcars$hp~mtcars$mpg,xlab='MPG',ylab='HP',col='green')
abline(lsfit(mtcars$mpg, mtcars$hp),col = "pink")
```



```
chris_model<-lm(formula =hp~mpg, data = mtcars )</pre>
summary(chris_model)
##
## Call:
## lm(formula = hp ~ mpg, data = mtcars)
## Residuals:
      Min
              1Q Median
##
                             3Q
## -59.26 -28.93 -13.45 25.65 143.36
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
                 324.08
                              27.43 11.813 8.25e-13 ***
## (Intercept)
                  -8.83
                               1.31 -6.742 1.79e-07 ***
## mpg
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 43.95 on 30 degrees of freedom
## Multiple R-squared: 0.6024, Adjusted R-squared: 0.5892
## F-statistic: 45.46 on 1 and 30 DF, p-value: 1.788e-07
#This model has an accuracy of 0.6024.
#From the above models, it can be concluded that the estimation done by Chris is correct.
#2b) Build a model that uses the number of cylinders (cyl) and the mile per gallon (mpg) values of a car
to predict the car Horse Power (hp). Using this model, what is the estimated Horse Power of a car with 4
calendar and mpg of 22?
horse_power<-lm(hp~cyl+mpg,data = mtcars)</pre>
summary(horse_power)
##
## Call:
## lm(formula = hp ~ cyl + mpg, data = mtcars)
##
## Residuals:
##
      Min
              1Q Median
                             3Q
## -53.72 -22.18 -10.13 14.47 130.73
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
                             86.093
                                      0.628 0.53492
                 54.067
## (Intercept)
## cyl
                 23.979
                              7.346
                                      3.264 0.00281 **
## mpg
                 -2.775
                              2.177 -1.275 0.21253
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 38.22 on 29 degrees of freedom
## Multiple R-squared: 0.7093, Adjusted R-squared: 0.6892
```

F-statistic: 35.37 on 2 and 29 DF, p-value: 1.663e-08

```
estimate_horse_power<-predict(horse_power,data.frame(cyl=4,mpg=22))
estimate_horse_power</pre>
```

1 ## 88.93618

#The estimated Horse Power is 88.93618

#3. For this question, we are going to use BostonHousing dataset. The dataset is in 'mlbench' package, so we first need to install the package, call the library and load the dataset using the following commands.

```
#installing and running required packages
library('mlbench')

data(BostonHousing)
str(BostonHousing)
```

```
## 'data.frame':
                    506 obs. of 14 variables:
                    0.00632 0.02731 0.02729 0.03237 0.06905 ...
##
   $ crim
            : num
##
   $ zn
             : num 18 0 0 0 0 0 12.5 12.5 12.5 12.5 ...
                    2.31 7.07 7.07 2.18 2.18 2.18 7.87 7.87 7.87 7.87 ...
   $ indus : num
             : Factor w/ 2 levels "0","1": 1 1 1 1 1 1 1 1 1 1 ...
##
   $ chas
##
     nox
             : num
                    0.538 \ 0.469 \ 0.469 \ 0.458 \ 0.458 \ 0.524 \ 0.524 \ 0.524 \ 0.524 \ \dots
##
   $ rm
                    6.58 6.42 7.18 7 7.15 ...
             : num
   $ age
                    65.2 78.9 61.1 45.8 54.2 58.7 66.6 96.1 100 85.9 ...
##
             : num
                    4.09 4.97 4.97 6.06 6.06 ...
##
   $
     dis
             : num
                    1 2 2 3 3 3 5 5 5 5 ...
##
   $ rad
             : num
                    296 242 242 222 222 222 311 311 311 311 ...
##
   $ tax
             : num
                    15.3 17.8 17.8 18.7 18.7 15.2 15.2 15.2 15.2 ...
   $ ptratio: num
                    397 397 393 395 397 ...
##
   $ b
             : num
   $ lstat : num
##
                    4.98 9.14 4.03 2.94 5.33 ...
             : num 24 21.6 34.7 33.4 36.2 28.7 22.9 27.1 16.5 18.9 ...
```

#3a) Build a model to estimate the median value of owner-occupied homes (medv)based on the following variables: crime crate (crim), proportion of residential land zoned for lots over 25,000 sq.ft (zn), the local pupil-teacher ratio (ptratio) and weather the whether the tract bounds Chas River(chas). Is this an accurate model?

```
owner_occupied_homes <- lm(medv~crim+zn+ptratio+chas,data=BostonHousing)
summary(owner_occupied_homes)</pre>
```

```
##
## lm(formula = medv ~ crim + zn + ptratio + chas, data = BostonHousing)
##
## Residuals:
       Min
                1Q
                   Median
                                3Q
                                       Max
## -18.282 -4.505 -0.986
                             2.650
                                    32.656
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 49.91868
                           3.23497 15.431 < 2e-16 ***
```

```
-0.26018
                           0.04015
                                    -6.480 2.20e-10 ***
## crim
                0.07073
                                     4.570 6.14e-06 ***
## zn
                           0.01548
## ptratio
               -1.49367
                           0.17144
                                    -8.712 < 2e-16 ***
                                     3.496 0.000514 ***
                4.58393
                           1.31108
## chas1
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 7.388 on 501 degrees of freedom
## Multiple R-squared: 0.3599, Adjusted R-squared: 0.3547
## F-statistic: 70.41 on 4 and 501 DF, p-value: < 2.2e-16
```

#This model's accuracy is 0.3599. Therefore it is quite inaccurate.

#3b) Use the estimated coefficient to answer these questions?

#I. Imagine two houses that are identical in all aspects but one bounds the Chas River and the other does not. Which one is more expensive and by how much?

#Answer: Chas is a factorial variable. Number "1" stands for the ones that are bound to Chas River while "0" is for those who are not bound to the Chas River. From the data it can be concluded that the median value of owner-occupied homes is \$1,000, and the chas1 coefficient is 4.58393. When compared to a house that is not bound to the Chas River, the house that is bound to the Chas River will be more expensive by 4.58393 in \$1000.

#II. Imagine two houses that are identical in all aspects but in the neighborhood of one of them the pupil-teacher ratio is 15 and in the other one is 18. Which one is more expensive and by how much?

#Answer: The pupil to teacher ratio basically measures how many students a teacher is allotted, and the lower this number is, the more expensive the communities or the homes will be. Property prices decrease by 1.49367 for every unit that the ptratio rises (in thousands). If the ptratio is 15, the fall will be 15 * 1493.67=22405.05. If the ptratio is 18, for instance, the #decrease will be 18 * 1493.67=26886.06. Therefore, a 15 ptratio will cost \$4481.01 more, than an 18 ptratio.

#3c) Which of the variables are statistically important (i.e. related to the house price)?

#Answer: The p values for all variables are not equal to zero. #It is indicating that we can confidently reject the default #null hypothesis since there is no relationship between house price #and other variables in the model. Each variable is #therefore statistically significant.

#3d) Use the anova analysis and determine the order of importance of these four variables.

anova(owner_occupied_homes)

```
## Analysis of Variance Table
##
## Response: medv
##
                  Sum Sq Mean Sq F value
                  6440.8
                          6440.8 118.007 < 2.2e-16 ***
## crim
               1
## zn
               1
                  3554.3
                          3554.3 65.122 5.253e-15 ***
                  4709.5
                          4709.5 86.287 < 2.2e-16 ***
## ptratio
               1
## chas
               1
                   667.2
                           667.2 12.224 0.0005137 ***
## Residuals 501 27344.5
                            54.6
                  0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Signif. codes:
```

#Answer: In comparison to other variables, the crim variable is seen to explain a proportionally higher sum squared amount of variability. We can assume that adding the crim improved the model greatly. Residuals,

however, reveal that a significant portion of the variability is unaccounted for. #The rankings are as follows-#1) crime crate (crim) is 6440.8 #2) The ptratio (ptratio) is 4709.5 #3) zn is 3554.3 #4) Chas is 667.2