## assignment\_06\_KummarikuntaVidyasagar.R

## 12702

## 2020-10-09

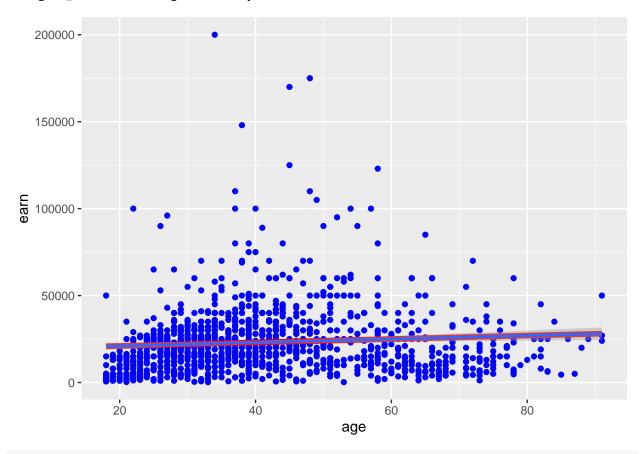
```
# Assignment: ASSIGNMENT 6
# Name: Kummarikunta, Vidyasagar
# Date: 2020-10-09
## Set the working directory to the root of your DSC 520 directory
setwd("/Users/12702/Desktop/MODatascience/DSC-520")
## Load the 'data/r4ds/heights.csv' to
heights_df <- read.csv("data/r4ds/heights.csv")
head(heights_df)
      earn
           height
                     sex ed age race
## 1 50000 74.42444 male 16 45 white
## 2 60000 65.53754 female 16 58 white
## 3 30000 63.62920 female 16 29 white
## 4 50000 63.10856 female 16 91 other
## 5 51000 63.40248 female 17 39 white
## 6 9000 64.39951 female 15 26 white
## Load the qqplot2 library
library(ggplot2)
## Fit a linear model using the 'age' variable as the predictor and 'earn' as the outcome
age_lm <- lm(formula = earn ~ age, data = heights_df)
## View the summary of your model using 'summary()'
summary(age_lm)
##
## Call:
## lm(formula = earn ~ age, data = heights_df)
##
## Residuals:
             1Q Median
## -25098 -12622 -3667
                         6883 177579
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 19041.53
                          1571.26 12.119 < 2e-16 ***
## age
                 99.41
                            35.46
                                   2.804 0.00514 **
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
```

```
## Residual standard error: 19420 on 1190 degrees of freedom
## Multiple R-squared: 0.006561, Adjusted R-squared: 0.005727
## F-statistic: 7.86 on 1 and 1190 DF, p-value: 0.005137

## Creating predictions using 'predict()'
age_predict_df <- data.frame(earn = predict(age_lm,heights_df ), age= heights_df$age)

## Plot the predictions against the original data
ggplot(data = heights_df, aes(y = earn, x = age)) +
geom_point(color='blue') +
geom_line(color='red',data = age_predict_df, aes(y= earn, x=age), size=2) + stat_smooth(method = lm)</pre>
```

## ## 'geom\_smooth()' using formula 'y ~ x'



```
mean_earn <- mean(heights_df$earn)
## Corrected Sum of Squares Total
sst <- sum((mean_earn - heights_df$earn)^2)
## Corrected Sum of Squares for Model
ssm <- sum((mean_earn - age_predict_df$earn)^2)
## Residuals
residuals <- heights_df$earn - age_predict_df$earn
## Sum of Squares for Error
sse <- sum(residuals^2)
## R Squared R^2 = SSM\SST
r_squared <- ssm/sst
## Number of observations</pre>
```

```
n <- nrow(heights_df)</pre>
## Number of regression parameters
p <- 2
## Corrected Degrees of Freedom for Model (p-1)
dfm \leftarrow (p-1)
## Degrees of Freedom for Error (n-p)
dfe \leftarrow (n-p)
## Corrected Degrees of Freedom Total: DFT = n - 1
dft \leftarrow (n-1)
## Mean of Squares for Model: MSM = SSM / DFM
msm \leftarrow ssm/dfm
## Mean of Squares for Error: MSE = SSE / DFE
mse <- sse/dfe</pre>
## Mean of Squares Total: MST = SST / DFT
mst <- sst/dft</pre>
## F Statistic F = MSM/MSE
f_score <- msm/mse</pre>
## Adjusted R Squared R2 = 1 - (1 - R2)(n - 1) / (n - p)
adjusted_r_squared <- 1-(1-r_squared)*(n-1)/(n-p)
\#\# Calculate the p-value from the F distribution
p_value <- pf(f_score, dfm, dft, lower.tail=F)</pre>
print(p_value)
```

## [1] 0.005136826