

assignment_06_BrownLincoln.R

x

2021-08-01

```
# Assignment: ASSIGNMENT 6
# Name: Brown, Lincoln
# Date: 2021-07-28

## Set the working directory to the root of your DSC 520 directory
setwd("/media/x/disk/School/DSC520/git/dsc520/")

## Load the `data/r4ds/heights.csv` to
heights_df <- read.csv("data/r4ds/heights.csv")

## Load the ggplot2 library
library(ggplot2)

## Fit a linear model using the `age` variable as the predictor and `earn` as the outcome
## lm([target variable] ~ [predictor variables], data = [data source])
age_lm <- lm(formula = "earn ~ age", data = heights_df)

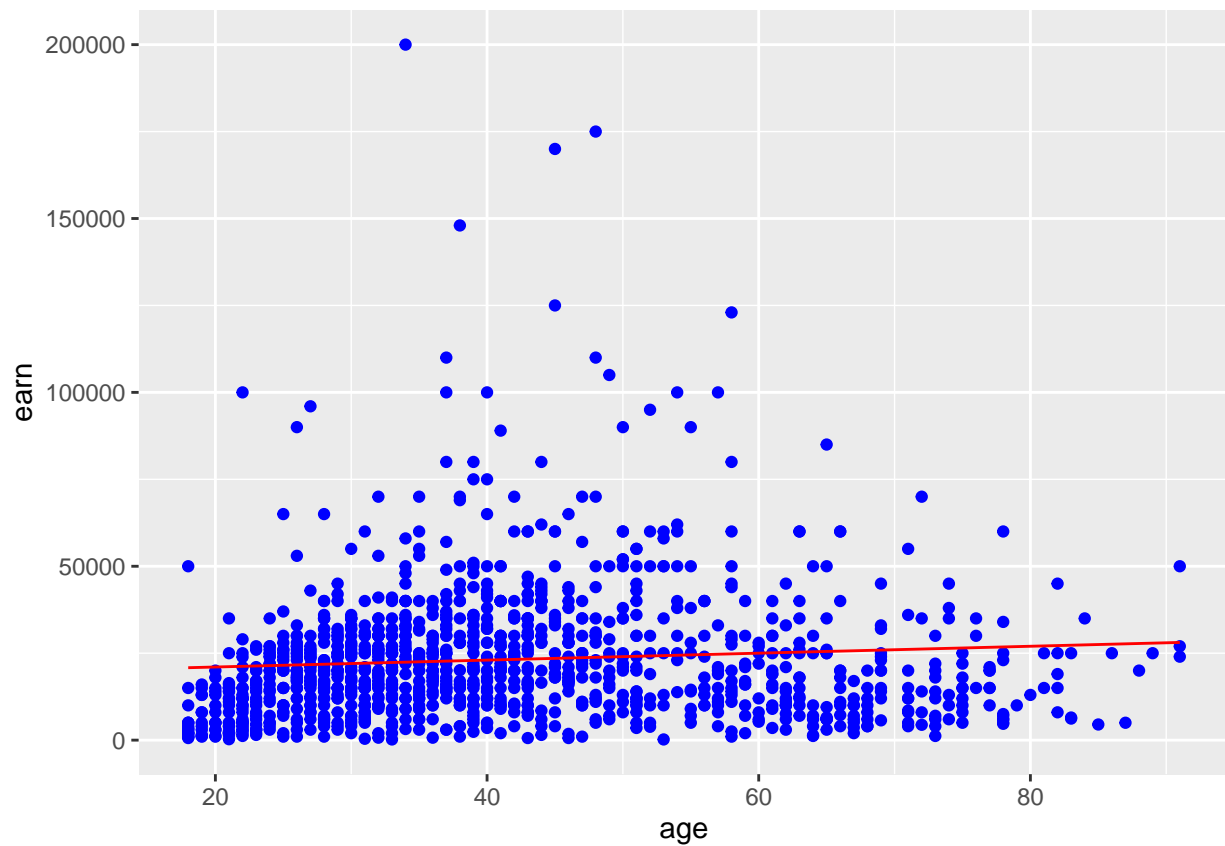
## View the summary of your model using `summary()`
summary(age_lm)

##
## Call:
## lm(formula = "earn ~ age", data = heights_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -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))
```



```
mean_earn <- mean(heights_df$earn)
mean_earn
```

```
## [1] 23154.77
```

```
## Corrected Sum of Squares Total
```

```
sst <- sum((mean_earn - heights_df$earn)^2)
sst
```

```
## [1] 451591883937
```

```
## Corrected Sum of Squares for Model
```

```
ssm <- sum((mean_earn - age_predict_df$earn)^2)
ssm
```

```
## [1] 2963111900
```

```
## Residuals
```

```
residuals <- heights_df$earn - age_predict_df$earn
```

```
## Sum of Squares for Error
```

```
sse <- sum(residuals^2)
sse
```

```
## [1] 448628772037
```

```
## R Squared  $R^2 = SSM/SST$ 
```

```
r_squared <- ssm/sst  
r_squared
```

```
## [1] 0.006561482
```

```
## Number of observations
```

```
n <- nrow(heights_df)
```

```
## Number of regression parameters
```

```
p <- 2
```

```
## Corrected Degrees of Freedom for Model (p-1)
```

```
dfm <- p-1
```

```
print(dfm)
```

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

```
## Degrees of Freedom for Error (n-p)
```

```
dfe <- n-p
```

```
print(dfe)
```

```
## [1] 1190
```

```
## Corrected Degrees of Freedom Total:  $DFT = n - 1$ 
```

```
dft <- n-1
```

```
print(dft)
```

```
## [1] 1191
```

```
## Mean of Squares for Model:  $MSM = SSM / DFM$ 
```

```
msm <- ssm / dfm
```

```
print(msm)
```

```
## [1] 2963111900
```

```
## Mean of Squares for Error:  $MSE = SSE / DFE$ 
```

```
mse <- sse / dfe
```

```
print(mse)
```

```
## [1] 376998968
```

```
## Mean of Squares Total:  $MST = SST / DFT$ 
```

```
mst <- sst / dft
```

```
print(mst)
```

```
## [1] 379170348
```

```
## F Statistic  $F = MSM/MSE$ 
```

```
f_score <- msm/mse
```

```
print(f_score)
```

```
## [1] 7.859735
```

```
## Adjusted R Squared  $R^2 = 1 - (1 - R^2)(n - 1) / (n - p)$ 
```

```
adjusted_r_squared <- 1-(1-r_squared)*(n-1) / (n-p)
```

```
print(adjusted_r_squared)
```

```
## [1] 0.005726659
```

```
## Calculate the p-value from the F distribution
```

```
p_value <- pf(f_score, dfm, dft, lower.tail=F)
```

```
print(p_value)
```

```
## [1] 0.005136826
```