

May 15, 2022

The results below are generated from an R script.

```
# Assignment: ASSIGNMENT 7
# Name: Jordan, Andrew
# Date: 2022-05-11

## Set the working directory to the root of your DSC 520 directory
setwd("/Users/Andrew/StatClass/dsc520/")

## Load the 'data/r4ds/heights.csv' to
heights_df <- read.csv("data/r4ds/heights.csv")
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
## 7	29000	61.65633	female	12	49	white
## 8	32000	72.69854	male	17	46	white
## 9	2000	72.03947	male	15	21	hispanic
## 10	27000	72.23493	male	12	26	white
## 11	6530	69.51215	male	16	65	white
## 12	30000	68.03161	male	11	34	white
## 13	12000	67.55693	male	12	27	white
## 14	12000	65.43059	female	12	51	white
## 15	22000	65.66285	female	16	35	white
## 16	17000	67.75877	male	12	58	white
## 17	40000	68.35184	female	14	29	white
## 18	44000	69.60957	male	13	44	white
## 19	7000	64.18457	female	12	55	black
## 20	53000	73.07461	male	13	35	black
## 21	5000	62.37553	female	13	51	white
## 22	14000	63.02393	female	14	21	white
## 23	5500	67.22990	male	14	22	white
## 24	40000	65.55111	female	12	41	white
## 25	34000	72.07965	male	12	45	white
## 26	10000	63.09113	female	12	35	black
## 27	27000	64.32355	female	16	60	white
## 28	50000	71.64285	male	16	38	white
## 29	41000	76.79309	male	16	33	white
## 30	15000	63.89391	female	14	25	white
## 31	25000	63.80262	female	12	33	white

##	32	75000	71.59223	male	17	39	white
##	33	27000	67.52196	male	17	31	white
##	34	12000	64.39435	female	12	26	white
##	35	7500	61.17822	female	14	78	white
##	36	30000	66.98388	female	14	31	black
##	37	21000	65.31646	female	12	57	white
##	38	27000	63.57419	female	14	26	white
##	39	3000	66.61100	female	15	65	white
##	40	25000	64.91176	female	12	30	white
##	41	24000	64.78968	female	12	41	white
##	42	32000	66.93769	female	18	29	white
##	43	10000	68.17281	female	17	30	white
##	44	11000	60.45066	female	12	21	hispanic
##	45	18700	64.79325	female	13	32	white
##	46	20000	61.81492	female	12	29	white
##	47	3500	71.57215	male	10	18	white
##	48	13000	67.31441	male	8	56	black
##	49	25000	69.89987	male	12	65	white
##	50	21000	69.76170	male	17	41	white
##	51	34000	67.74647	female	17	49	white
##	52	6000	60.19022	female	12	65	white
##	53	17000	71.00650	male	12	28	white
##	54	35000	71.16680	male	12	32	white
##	55	4000	72.73563	male	13	18	white
##	56	14000	68.13822	female	14	55	white
##	57	10000	66.37981	female	12	57	white
##	58	25000	69.23278	male	16	29	white
##	59	16000	63.27394	female	14	27	white
##	60	16000	61.82776	male	14	28	hispanic
##	61	16500	64.22121	female	14	43	white
##	62	4000	63.84127	female	9	68	white
##	63	3840	66.97477	female	9	52	white
##	64	22000	71.45149	male	12	39	white
##	65	200	59.61265	female	16	53	white
##	66	26000	65.79939	female	16	27	white
##	67	2500	66.45804	female	15	21	white
##	68	17000	64.60288	female	14	39	white
##	69	8000	70.44048	female	13	22	white
##	70	12000	65.92281	female	13	68	white
##	71	10000	61.85683	female	12	47	white
##	72	10000	65.78444	female	15	67	white
##	73	15000	71.83128	male	12	39	white
##	74	2400	67.04533	female	8	39	hispanic
##	75	30000	68.30551	male	12	32	hispanic
##	76	30000	70.02546	male	12	33	white
##	77	10000	61.81039	female	12	38	white
##	78	5000	62.95107	female	13	26	white
##	79	12000	65.82114	female	13	63	white
##	80	20000	70.39755	female	10	61	white
##	81	20000	68.37778	female	12	36	white
##	82	20000	69.93270	male	14	23	white
##	83	1200	66.17181	female	12	20	white
##	84	700	68.45636	female	16	32	white
##	85	20000	69.90386	male	16	27	white

##	86	10000	61.14966	female	12	22	hispanic
##	87	30000	63.36335	female	12	73	white
##	88	40000	64.14708	female	14	56	white
##	89	25000	67.31839	male	12	89	white
##	90	10000	60.67494	female	17	79	white
##	91	60000	68.84090	female	18	63	white
##	92	18000	67.68273	female	12	66	white
##	93	16040	64.49677	female	12	33	white
##	94	15000	66.81240	female	14	30	black
##	95	10000	68.74644	male	17	23	white
##	96	33000	67.06765	female	13	43	white
##	97	18000	68.13799	female	12	30	white
##	98	15000	63.34290	female	12	37	white
##	99	21000	71.38667	male	12	22	white
##	100	21000	63.98834	female	17	43	black
##	101	37000	68.48639	male	11	37	white
##	102	38000	67.51614	female	17	44	white
##	103	17000	65.60084	female	14	43	hispanic
##	104	32000	76.80019	male	16	30	white
##	105	27500	67.10538	female	12	58	white
##	106	16500	62.15164	female	12	44	white
##	107	25000	66.86762	female	18	35	white
##	108	27000	61.04220	female	18	43	white
##	109	5000	64.12329	female	12	28	white
##	110	70000	61.54482	female	16	38	white
##	111	5000	62.55624	female	12	40	white
##	112	5000	68.16377	male	16	24	white
##	113	20000	63.65513	female	15	26	white
##	114	4000	72.37352	male	15	21	white
##	115	60000	64.14708	female	16	35	white
##	116	5000	61.32670	female	13	31	white
##	117	30000	74.36640	male	12	38	white
##	118	70000	70.21016	male	14	35	white
##	119	50000	71.10619	male	16	41	white
##	120	44000	62.59484	female	12	39	white
##	121	30000	64.05496	female	14	43	white
##	122	10000	61.57362	female	16	40	white
##	123	23000	70.48020	female	17	42	white
##	124	45000	71.18591	male	17	62	white
##	125	15000	71.43364	male	14	31	white
##	126	4000	70.22885	female	14	71	white
##	127	17000	67.28086	male	14	31	white
##	128	30000	63.75869	female	12	32	white
##	129	27500	67.08652	female	12	30	white
##	130	5688	61.67960	female	8	69	white
##	131	18000	62.28600	female	13	56	hispanic
##	132	43000	68.29248	male	13	44	black
##	133	32000	61.58948	female	14	44	black
##	134	10000	68.41774	female	18	56	black
##	135	60000	73.99126	male	13	45	white
##	136	21000	67.56107	female	12	50	other
##	137	2400	62.33793	female	16	22	white
##	138	1000	66.24001	female	15	28	white
##	139	27000	68.09847	male	12	27	white

```

## 140 6600 59.77087 female 14 28 hispanic
## 141 16000 68.06338 male 8 43 white
## 142 90000 71.68015 male 12 26 white
## 143 8000 66.35971 female 12 42 white
## 144 20000 68.35626 male 10 32 white
## 145 15000 68.45654 female 12 18 white
## 146 12000 68.78610 female 12 60 white
## 147 24000 64.10224 female 16 46 white
## 148 20000 65.11349 female 14 39 white
## 149 19000 60.64919 female 12 46 white
## 150 10000 72.12570 male 12 49 white
## 151 40000 65.51073 female 16 34 white
## 152 25000 67.93190 male 14 64 white
## 153 25000 70.44492 male 12 24 white
## 154 25000 71.36585 male 14 32 white
## 155 19000 71.12507 male 16 61 white
## 156 44000 68.16014 male 16 48 white
## 157 15000 60.11333 female 14 49 white
## 158 17000 62.78820 female 12 36 white
## 159 24000 68.07772 male 12 56 white
## 160 23000 64.05084 female 12 37 white
## 161 13000 69.71580 male 12 74 white
## 162 65000 68.22067 male 16 46 white
## 163 7000 60.88386 female 12 63 white
## 164 40000 68.40754 male 18 63 white
## 165 15000 66.00198 female 17 43 white
## 166 20000 69.79789 male 16 25 white
## [ reached 'max' / getOption("max.print") -- omitted 1026 rows ]

# Fit a linear model
earn_lm <- lm(earn ~ age + ed + race + height + sex, data=heights_df)

# View the summary of your model
summary(earn_lm)

##
## Call:
## lm(formula = earn ~ age + ed + race + height + sex, data = heights_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -39423  -9827  -2208   6157 158723
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -41478.4    12409.4   -3.342  0.000856 ***
## age             178.3       32.2    5.537  3.78e-08 ***
## ed             2768.4       209.9   13.190 < 2e-16 ***
## racehispanic  -1414.3     2685.2   -0.527  0.598507
## raceother      371.0     3837.0    0.097  0.922983
## racewhite     2432.5     1723.9    1.411  0.158489
## height         202.5       185.6    1.091  0.275420
## sexmale       10325.6     1424.5    7.249  7.57e-13 ***
## ---

```

```

## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 17250 on 1184 degrees of freedom
## Multiple R-squared:  0.2199, Adjusted R-squared:  0.2153
## F-statistic: 47.68 on 7 and 1184 DF,  p-value: < 2.2e-16

predicted_df <- data.frame(
  earn = predict(earn_lm,heights_df),
  ed=heights_df$ed, race=heights_df$race, height=heights_df$height,
  age=heights_df$age, sex=heights_df$sex
)

## Compute deviation (i.e. residuals)
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 - predicted_df$earn)^2)
## Residuals
residuals <- heights_df$earn - predicted_df$earn
## Sum of Squares for Error
sse <- sum(residuals^2)
## R Squared
r_squared <- ssm/sst

## Number of observations
n <- sum(complete.cases(heights_df))
n

## [1] 1192

## Number of regression paramaters
p <- 8
## Corrected Degrees of Freedom for Model
dfm <- p-1
## Degrees of Freedom for Error
dfe <- n-p
## Corrected Degrees of Freedom Total:  DFT = n - 1
dft <- n-1

## Mean of Squares for Model:  MSM = SSM / DFM
msm <- ssm/dfm
## Mean of Squares for Error:  MSE = SSE / DFE
mse <- sse/dfe
## Mean of Squares Total:  MST = SST / DFT
mst <- sst/dft
## F Statistic
f_score <- msm/mse

## Adjusted R Squared  $R^2 = 1 - (1 - R^2)(n - 1) / (n - p)$ 
adjusted_r_squared <- 1-(1-r_squared)*(n-1)/(n-p)

```

The R session information (including the OS info, R version and all packages used):

```

sessionInfo()

## R version 4.2.0 (2022-04-22 ucrt)
## Platform: x86_64-w64-mingw32/x64 (64-bit)
## Running under: Windows 10 x64 (build 19044)
##
## Matrix products: default
##
## locale:
## [1] LC_COLLATE=English_United States.utf8  LC_CTYPE=English_United States.utf8
## [3] LC_MONETARY=English_United States.utf8 LC_NUMERIC=C
## [5] LC_TIME=English_United States.utf8
##
## attached base packages:
## [1] stats      graphics  grDevices  utils      datasets  methods   base
##
## other attached packages:
## [1] car_3.0-13      carData_3.0-5 lm.beta_1.6-2 ggplot2_3.3.6 readxl_1.4.0
##
## loaded via a namespace (and not attached):
## [1] highr_0.9      cellranger_1.1.0 pillar_1.7.0    compiler_4.2.0  tools_4.2.0
## [6] digest_0.6.29  evaluate_0.15    lifecycle_1.0.1 tibble_3.1.7    gtable_0.3.0
## [11] pkgconfig_2.0.3 rlang_1.0.2      cli_3.3.0       yaml_2.3.5      xfun_0.30
## [16] fastmap_1.1.0  stringr_1.4.0    withr_2.5.0     dplyr_1.0.9     knitr_1.39
## [21] generics_0.1.2 vctrs_0.4.1      grid_4.2.0      tidyselect_1.1.2 glue_1.6.2
## [26] R6_2.5.1       fansi_1.0.3      rmarkdown_2.14  farver_2.1.0    purrr_0.3.4
## [31] magrittr_2.0.3 scales_1.2.0     ellipsis_0.3.2  htmltools_0.5.2 abind_1.4-5
## [36] colorspace_2.0-3 labeling_0.4.2    tinytex_0.38    utf8_1.2.2      stringi_1.7.6
## [41] munsell_0.5.0  crayon_1.5.1

Sys.time()

## [1] "2022-05-15 13:02:24 EDT"

```