Homework 1

Juwon Lee, Economics and Statistics, UCLA

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tinytex::install tinytex()

1. Review on simple linear regression.

- (1) B. They have moderate negative linear relationship.
- (2)

Correlation Coefficient $r = \sqrt{R^2} = \sqrt{0.4117} \approx 0.64164$

(3) This is

$$\hat{\beta}_1 = r * \frac{s_Y}{s_X} = 0.64164 * \frac{5}{10.84} \approx 0.29596$$

$$\hat{\beta}_0 = \bar{Y} - \hat{\beta}_1 * \bar{X} = 20.16 - \hat{\beta}_1 * 20.97 \approx 13.95375$$

It means that we can make a 'model' to estimate Y such that $Y = \hat{\beta}_0 + \hat{\beta}_1 X$.

- (4) C. About 41% of the total variability in strength of deltoid muscle is explained by the model.
- (5) C. It does not have any apparent pattern and it implies the good fit of the model.
- (6) It means that X = 35, so that

$$Y = \hat{\beta}_0 + \hat{\beta}_1 X \approx 13.95375 + 0.29596 * 35 = 24.31230$$

2. Review on Hypothesis testing.

- (1) $H_0: \mu = 5.7$ vs $H_1: \mu < 5.7$.
- (2) Because n = 36 > 30, we can use z-statistic,

$$z = \frac{5.1 - 5.7}{1.4/\sqrt{36}} \approx -2.57143$$

(3) $P(z \le -2.57143) \approx 0.00506$.

[1] 0.005063995

(4) p-value $\approx 0.00506 < 0.05$, so we can ignore the null.

Thus, we can't say that the number of introverts are 5.7.

(5)
$$(5.7 - z_{0.025} * \frac{1.4}{\sqrt{36}}, 5.7 + z_{0.025} * \frac{1.4}{\sqrt{36}}) \approx (5.24268, 6.15733).$$

(6) Thus, if the sample mean is included in this interval, then we can say that the null hypothesis can be rejected. Thus, because $\bar{X} = 5.1$, we can reject the null.

3. Review on R programming

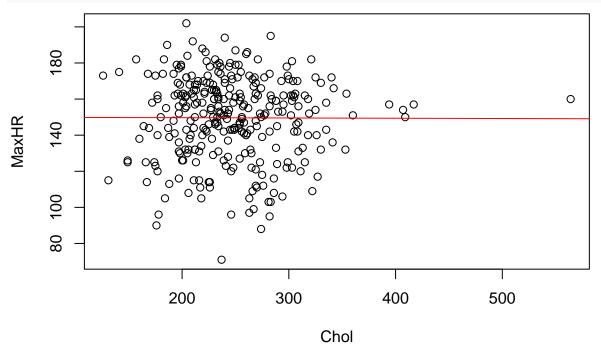
First of all, we can make a dataset by the csv file.

```
heart <- read.csv("/Users/user/Desktop/Yonsei/Junior/3-2/Introduction to Data Analysis and Regression/H
heart1 \leftarrow heart[c(6,9)]
head(heart1)
##
     Chol MaxHR
## 1 233
            150
## 2 286
            108
## 3 229
            129
## 4 250
            187
      204
            172
## 5
## 6
      236
            178
(1)
MaxHR <- as.vector(heart1[2])</pre>
Chol <- as.vector(heart1[1])</pre>
data_lm <- lm(MaxHR~Chol, data=heart1)</pre>
{\tt data\_lm}
##
## lm(formula = MaxHR ~ Chol, data = heart1)
## Coefficients:
## (Intercept)
                        Chol
## 149.981292 -0.001516
```

```
summary(data_lm)
```

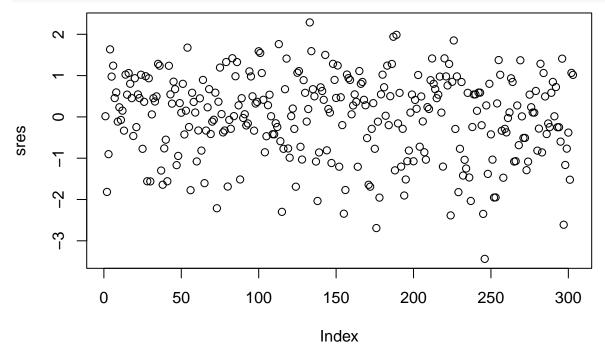
```
##
## Call:
## lm(formula = MaxHR ~ Chol, data = heart1)
##
## Residuals:
##
       Min
                1Q Median
                                 3Q
                                        Max
                     3.375 16.412 52.328
## -78.622 -16.079
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept) 149.981292
                            6.418400
                                        23.37
                                                <2e-16 ***
                            0.025465
                                                 0.953
## Chol
                -0.001516
                                        -0.06
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
\mbox{\tt \#\#} Residual standard error: 22.91 on 301 degrees of freedom
## Multiple R-squared: 1.178e-05, Adjusted R-squared: -0.00331
## F-statistic: 0.003545 on 1 and 301 DF, p-value: 0.9526
Thus, y = 149.981292 - 0.001516x.
(2)
```

```
plot(heart1)
abline(149.981292, -0.001516, col='red')
```



(3) The R-squared in summary is 1.178e-05. We have to use standardized residuals, so that

sres <- data_lm\$residuals/sqrt(var(data_lm\$residuals)) plot(sres)</pre>



If the residual plots does not show any pattern, then the model is supposed to be a good fit. However, in this case, the residual plot is almost the same as the scatter plot, suggesting that the model provides little information. Therefore, it is a poor fit.

4. Pre-Course Survey

I have completed the Pre-Course Survey. Thank you.