

# Pulse Rate Study

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```
library(tidyverse)
library(broom)
```

## Question 1

(a)

```
url <- "http://ritsokiguess.site/STAC32/pulsemarch.csv"
data <- read_csv(url)
```

```
##
## -- Column specification -----
## cols(
##   Sex = col_character(),
##   Before = col_double(),
##   After = col_double()
## )
data
```

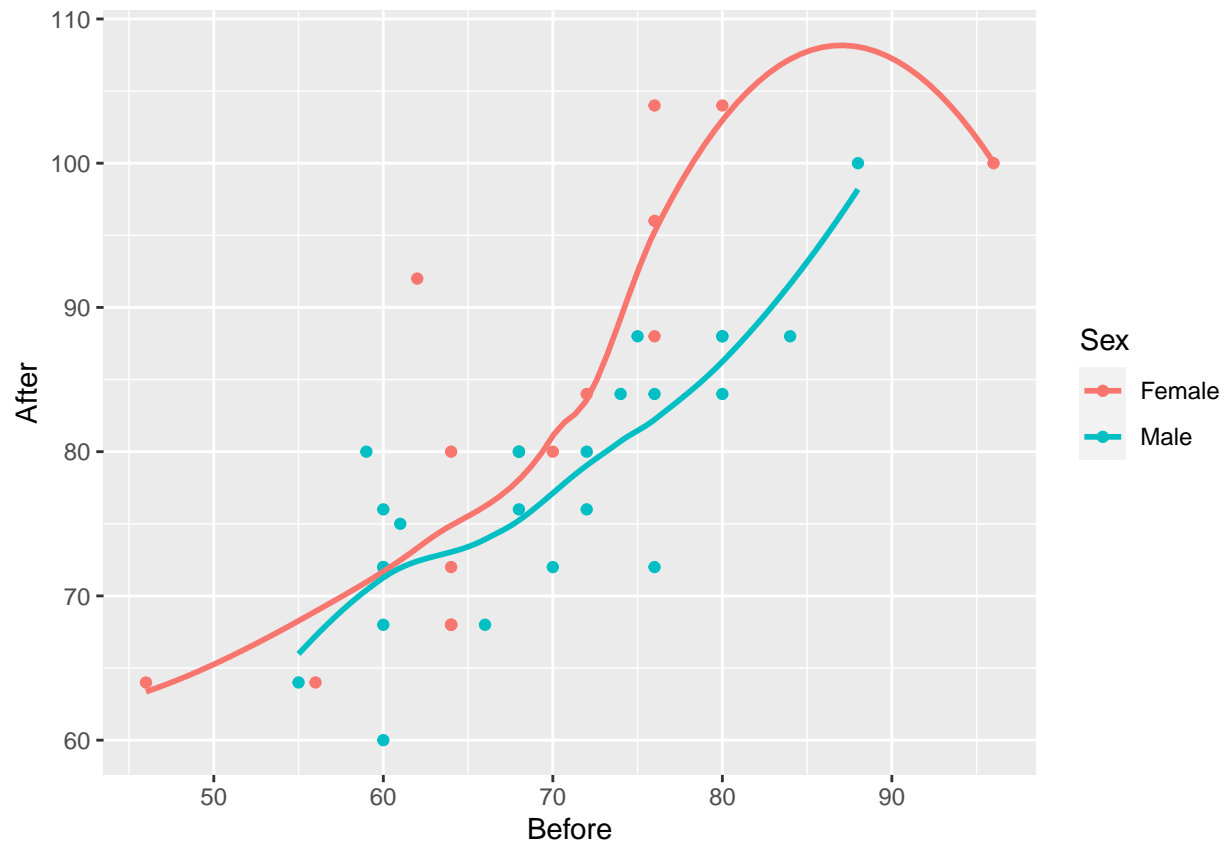
```
## # A tibble: 40 x 3
##   Sex      Before After
##   <chr>    <dbl> <dbl>
## 1 Female      72     84
## 2 Male       60     72
## 3 Female      68     80
## 4 Male       70     72
## 5 Male       68     80
## 6 Male       61     75
## 7 Male       80     84
## 8 Male       72     76
## 9 Female      64     80
## 10 Female     62     92
## # ... with 30 more rows
```

- use “read\_csv” because it’s a csv file.

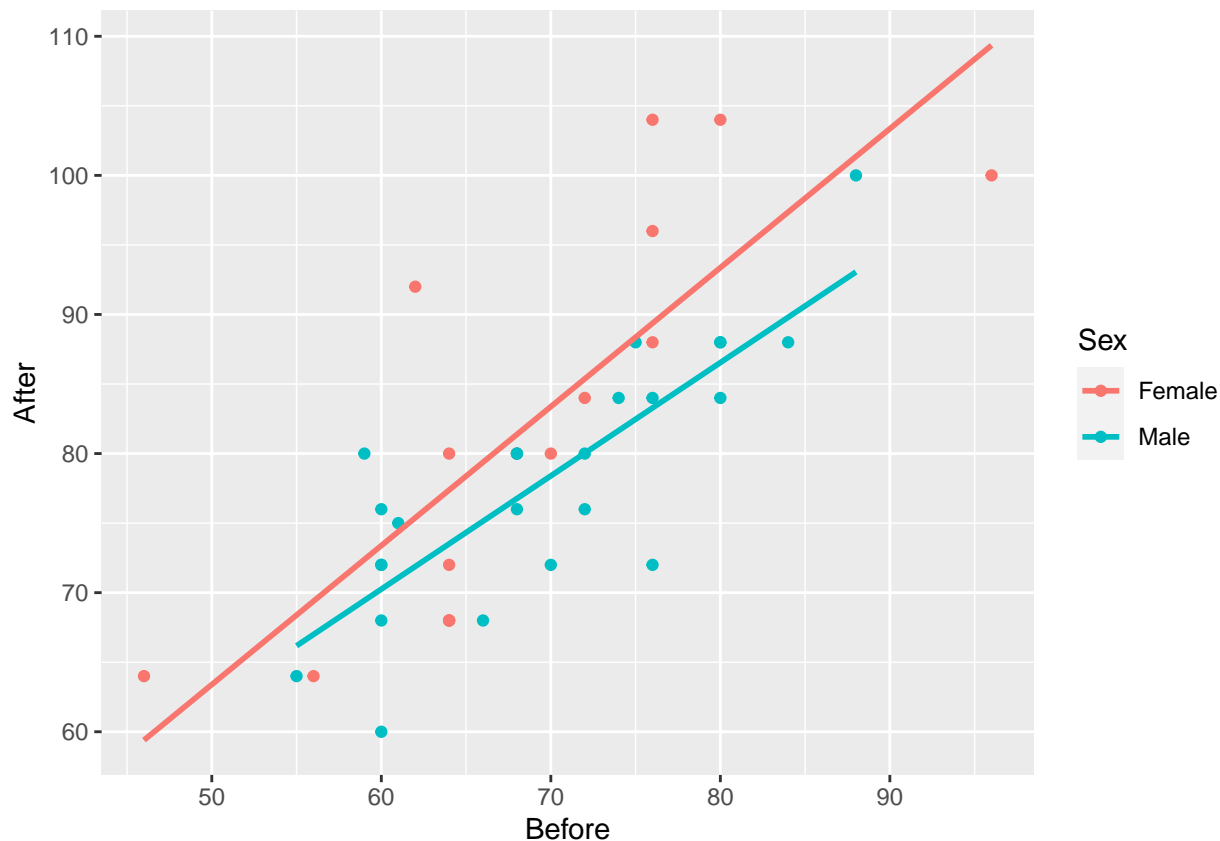
(b)

```
ggplot(data, aes(x = Before, y = After, colour = Sex)) + geom_point() + geom_smooth(se = F)

## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```



```
ggplot(data, aes(x = Before, y = After, colour = Sex)) + geom_point() + geom_smooth(method = "lm", se =
## `geom_smooth()` using formula 'y ~ x'
```



- The X variable is Before and Y variable is After because the Before pulse is our explanatory variable and the After pulse is our response variable.
- Use blue to represent males and red to represent females.
- First, we don't assume linear, but we can see from the first graph that both males and females can pass having a linear relationship between before and after.
- Add a linear regression line in the second graph.

(c)

- As seen from the graph, females generally have a higher after pulse rate than males do.
- Also, for most points all the graph—both males and females, the after pulse rates are higher than the before pulse rates.
- From the linear regression line, the slope of the female pulse rate is slightly larger than the male's. This means that the rate of pulse increase is a little higher for females.

(d)

```
fit <- lm(After ~ Before + Sex, data = data)
summary(fit)
```

```
##
## Call:
## lm(formula = After ~ Before + Sex, data = data)
##
```

```
## Residuals:
##      Min       1Q   Median       3Q      Max
## -11.8653  -4.6319  -0.4271   3.3856  16.0047
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  19.8003     7.9217   2.499  0.0170 *
## Before       0.9064     0.1127   8.046  1.2e-09 ***
## SexMale     -4.8191     2.2358  -2.155  0.0377 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.918 on 37 degrees of freedom
## Multiple R-squared:  0.6468, Adjusted R-squared:  0.6277
## F-statistic: 33.87 on 2 and 37 DF,  p-value: 4.355e-09

glance(fit)

## # A tibble: 1 x 12
##   r.squared adj.r.squared sigma statistic p.value    df logLik   AIC   BIC
##   <dbl>      <dbl> <dbl>      <dbl>   <dbl> <dbl> <dbl> <dbl> <dbl>
## 1    0.647        0.628  6.92        33.9 4.36e-9     2  -133.  273.  280.
## # ... with 3 more variables: deviance <dbl>, df.residual <int>, nobs <int>

tidy(fit)

## # A tibble: 3 x 5
##   term          estimate std.error statistic    p.value
##   <chr>          <dbl>    <dbl>      <dbl>    <dbl>
## 1 (Intercept)    19.8      7.92        2.50 0.0170
## 2 Before         0.906     0.113       8.05 0.00000000120
## 3 SexMale       -4.82     2.24       -2.16 0.0377
```

(e)

Under the linear regression model:

- The “SexMale” tells me the after pulse of males is about 4.8191 lower than that for females at the same before pulse.
- The slope of “Before” says the increasing before pulse by 1 increases the after pulse by about 0.9064 for both males and females.
- r squared is 0.6467726, which is not too terrible.
- The second slope does surprise me because I was expecting the slope to be greater than 1 since the after pulse should increase at a faster rate as the before pulse increases.

(f)

- As seen from the output, the estimate value of SexMale is -4.8191363.
- The “SexMale” tells me the after pulse of males is about 4.8191 lower than that for females at the same before pulse.

## Question 2

(a)

```
cv <- function(x){  
  sd(x)/mean(x)  
}
```

(b)

```
x <- 1:5  
cv(x)
```

```
## [1] 0.5270463
```

- Hence, the coefficient of variation of the set of integers 1 through 5 is 0.5270463.

(c)

```
x <- c(-2.8, -1.8, -0.8, 1.2, 4.2)  
cv(x)
```

```
## [1] 6.248491e+16
```

- The coefficient is  $6.2484909 \times 10^{16}$ .
- It doesn't make sense because the mean of x is 0. But we can't divide by 0 because that'd be undefined.

```
mean(x)
```

```
## [1] 4.440892e-17
```

(d)

```
cv <- function(x){  
  stopifnot(x>=0)  
  sd(x)/mean(x)  
}
```

- Use “stopifnot” to give an error message if the vector is negative.
- Test the function using our previous inputs:

```
x <- 1:5  
cv(x)
```

```
## [1] 0.5270463
```

```
x <- c(-2.8, -1.8, -0.8, 1.2, 4.2)  
cv(x)
```

```
## Error in cv(x): x >= 0 are not all TRUE
```

- Notice the first input outputs the same value we got from part (b)
- But the second input outputs error message since there are negative numbers, which is exactly what we wanted.