

Lab 2: Linear Regression

An Island Never Cries

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Exercise 1

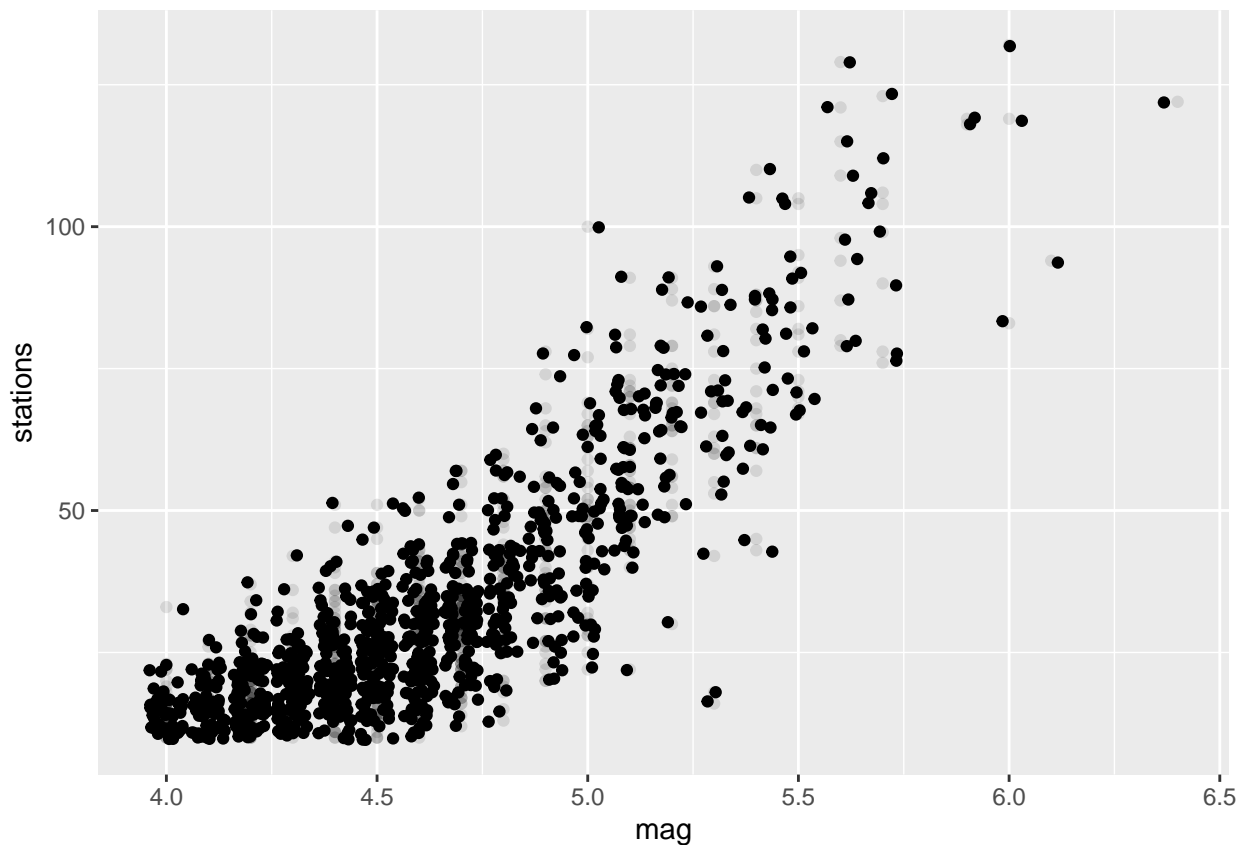
```
library(MASS)
library("ggplot2")
library("tidyverse")

## -- Attaching packages ----- tidyverse 1.2.1 --
## v tibble  2.1.1      v purrr   0.3.2
## v tidyr   0.8.3      v dplyr   0.8.0.1
## v readr   1.3.1      v stringr 1.4.0
## v tibble  2.1.1      v forcats 0.4.0

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## x dplyr::select() masks MASS::select()

data(quakes)

ggplot(quakes, aes(mag, stations)) + geom_point(alpha = 0.1) + geom_jitter()
```



There is a strong positive linear relationship between the magnitude of an earthquake and the number of stations that detect the earthquake. In general, as the magnitude of the earthquake increases, the number of reporting stations increases as well. There is a strong positive linear relationship between the magnitude of an earthquake and the number of stations that detect the earthquake. In general, as the magnitude of the earthquake increases, the number of reporting stations increases as well.

Exercise 2

```
mean(quakes$stations)
```

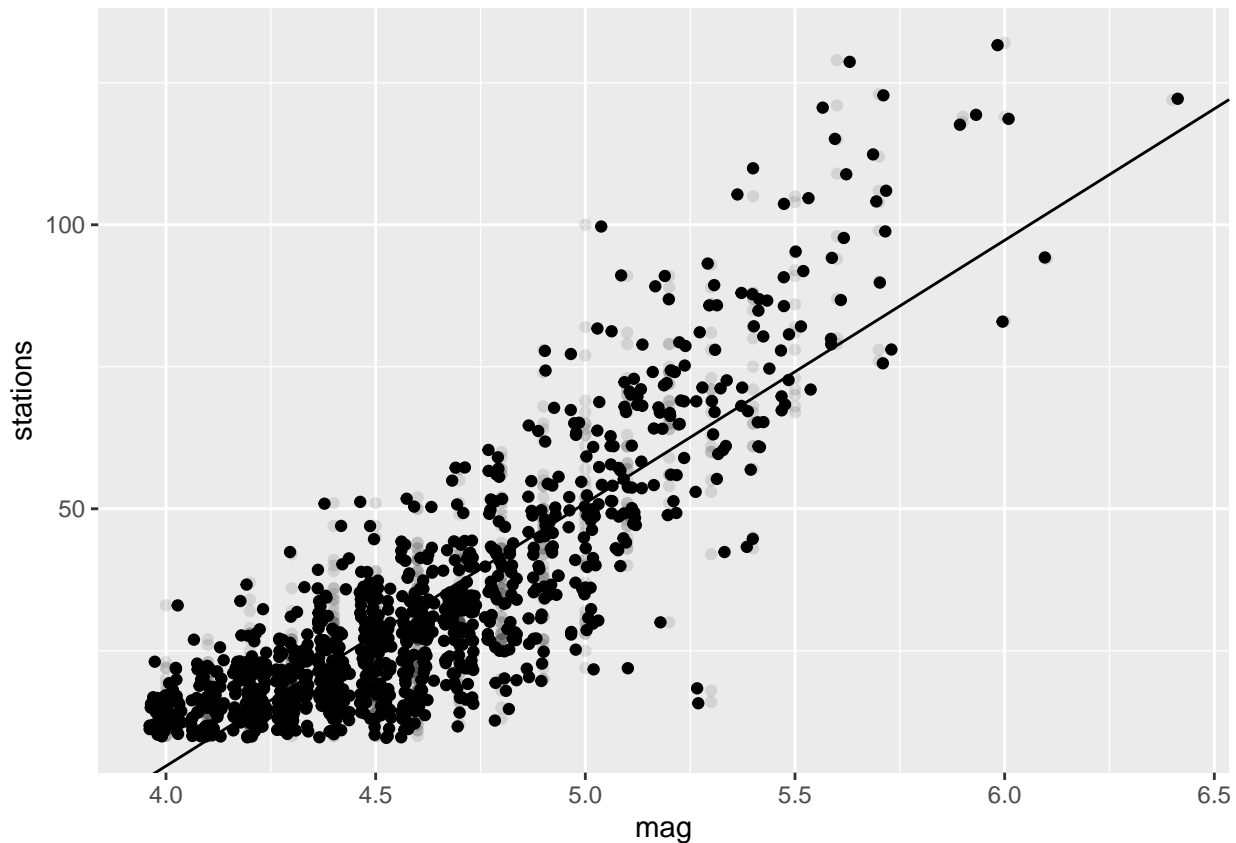
```
## [1] 33.418
```

The slope would be zero. It follows that the intercept would be 33.418, the mean of Y, since $\hat{B}_0 = \bar{y} - \hat{B}_1 \bar{x}$

Exercise 3

```
m1 <- lm(stations ~ mag, data = quakes)
```

```
ggplot(quakes, aes(mag, stations)) + geom_point(alpha = 0.1) + geom_abline(intercept = m1$coef[1], slope = m1$coef[2])
```



```
summary(m1)
```

```
##
## Call:
## lm(formula = stations ~ mag, data = quakes)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -48.871  -7.102  -0.474   6.783  50.244
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -180.4243     4.1899  -43.06  <2e-16 ***
## mag          46.2822     0.9034   51.23  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 11.5 on 998 degrees of freedom
## Multiple R-squared:  0.7245, Adjusted R-squared:  0.7242
## F-statistic: 2625 on 1 and 998 DF, p-value: < 2.2e-16
```

The slope is 46.2822 and the intercept is -180.4243. The model predicts that as the magnitude of an earthquake increases by one unit on the Richter Magnitude scale, the number of stations that detect the earthquake will increase by approximately 46.2822. The model also predicts that if the magnitude of an earthquake on the Richter Magnitude scale is 0, the number of stations that can detect it will be -180.4243. That is to say, no stations will detect an earthquake or event that registers as zero on the Richter scale.

Exercise 4

```
r <- cor(quakes$stations, quakes$mag)

y_diff <- quakes$stations - mean(quakes$stations)
y_diff_squared <- (y_diff)^2
std_dev_y <- sqrt(sum(y_diff_squared/999))
std_dev_y
```

```
## [1] 21.90039
```

```
x_diff <- quakes$mag - mean(quakes$mag)
x_diff_squared <- (x_diff)^2
std_dev_x <- sqrt(sum(x_diff_squared/999))
std_dev_x
```

```
## [1] 0.402773
```

```
slope <- r*(std_dev_y/std_dev_x)
slope
```

```
## [1] 46.28221
```

Using the equation

$$\hat{\beta}_1 = r * (s_y/s_x)$$

, I calculated a slope of 46.28221 between the magnitude of the earthquakes and the number of reporting stations. This corresponds with the slope computed by the fitted linear model.

Exercise 5

```
summary(m1)$coef[[2]] + qt(1-(0.05/2), 998) * summary(m1)$coef[[4]]
```

```
## [1] 48.05498
```

```
summary(m1)$coef[[2]] - qt(1-(0.05/2), 998) * summary(m1)$coef[[4]]
```

```
## [1] 44.50944
```

```
confint(m1, 'mag', level = 0.95)
```

```
##          2.5 %    97.5 %
## mag 44.50944 48.05498
```

We are 95% confident that the true slope between the magnitude of the earthquakes and the number of stations that detect them lies between 44.50944 and 48.05498.

Exercise 6

```
predict(m1, data.frame(mag = 7.0))
```

```
##          1
## 143.5511
```

I predict that around 144 (143.5511) stations would be able to detect an earthquake with a magnitude of 7.0.

Exercise 7

Question 1 - Data description Question 2 - Inference Question 3 - Inference Question 4 - Inference Question 5 - Inference Question 6 - Prediction

SIMULATION

Exercise 9

```
dat <- seq(from = 4.0, to = 6.0, by = 0.1)
sim_dat <- sample(dat, size = 1000, replace = TRUE)
sim_dat
```

```
##      [1] 5.6 4.6 4.1 4.0 4.3 5.3 4.0 6.0 4.8 4.3 4.3 5.2 5.9 5.4 6.0 4.4 4.5
##     [18] 4.0 5.2 4.5 5.3 5.0 4.7 5.4 4.3 4.7 4.2 5.0 4.1 4.8 5.1 5.7 5.2 5.0
##     [35] 5.1 4.5 4.6 4.3 5.1 5.0 5.3 5.9 4.1 6.0 5.0 4.1 5.1 4.0 4.6 5.8 4.0
##     [52] 5.4 5.9 5.2 5.2 4.4 4.0 4.6 4.9 4.6 4.5 5.3 4.8 4.5 4.0 4.5 5.2 5.0
##     [69] 5.4 4.4 4.1 5.5 5.2 4.7 5.5 5.0 4.6 5.5 4.9 4.5 4.9 4.6 4.2 4.5 5.0
##     [86] 4.9 5.8 5.5 5.4 5.6 5.7 5.6 5.0 5.2 5.3 5.7 5.6 5.8 4.7 5.4 5.8 4.2
##    [103] 5.3 5.1 4.2 4.4 4.9 5.0 5.0 5.4 5.1 4.1 4.4 5.4 5.2 4.8 5.3 4.3 4.7
##    [120] 4.8 5.9 5.1 5.0 4.7 5.0 5.5 5.2 4.4 4.8 4.0 5.2 4.1 4.3 4.1 5.4 4.0
##    [137] 4.4 5.8 4.1 4.4 4.5 4.4 5.4 5.9 5.1 4.9 4.3 5.8 4.5 4.3 4.8 5.1 5.3
##    [154] 4.0 5.9 5.8 4.8 5.9 6.0 4.0 4.8 4.3 5.1 5.9 4.2 5.1 4.2 4.0 4.8 5.7
##    [171] 5.6 4.3 5.3 4.2 4.0 4.2 5.7 5.2 5.7 5.6 4.8 5.8 5.0 5.7 5.3 5.5 4.2
##    [188] 4.4 5.1 5.9 4.4 5.8 4.9 6.0 5.9 5.8 5.3 5.5 5.2 4.9 4.2 6.0 4.5 4.5
##    [205] 5.9 5.1 5.3 5.3 6.0 4.8 5.4 4.4 5.2 5.3 5.1 4.4 4.9 4.1 4.0 5.9 5.8
##    [222] 4.6 5.5 5.2 4.4 5.0 5.9 5.8 4.0 4.6 5.1 6.0 4.7 4.3 5.6 4.3 4.2 5.3
##    [239] 5.9 4.6 5.1 4.2 4.2 4.3 4.9 4.7 5.0 4.1 4.0 5.5 4.8 4.9 5.9 4.4 5.0
##    [256] 5.3 4.5 5.3 5.2 4.1 4.1 4.8 5.0 5.7 5.9 5.2 5.3 5.7 5.9 5.2 4.2 4.4
##    [273] 4.2 4.0 4.9 4.7 5.9 4.6 5.9 4.9 5.0 4.0 5.6 5.3 4.4 5.2 5.3 5.0 5.6
##    [290] 5.9 5.7 5.7 4.2 4.2 5.9 5.4 4.5 4.8 5.2 5.1 4.4 5.1 5.2 4.9 5.6 5.9
##    [307] 5.3 4.1 5.8 4.4 4.2 5.8 4.2 4.2 4.3 4.1 5.9 4.9 4.9 4.1 5.4 4.7 4.2
##    [324] 4.4 4.0 5.9 5.7 4.3 5.4 5.7 4.7 4.0 6.0 5.7 4.0 5.5 5.6 5.5 4.3 4.5
##    [341] 4.9 5.2 4.4 5.7 5.2 5.5 5.1 4.9 4.4 5.6 5.7 4.8 5.4 4.1 4.4 4.8 5.9
##    [358] 4.5 4.7 4.4 4.5 4.9 5.1 5.7 4.0 5.6 4.5 5.7 5.3 4.1 5.5 5.5 4.1 5.4
##    [375] 5.9 4.8 4.2 4.1 4.8 5.8 5.4 4.8 5.3 4.4 5.4 4.5 4.1 5.3 4.9 4.6 5.6
##    [392] 4.3 5.3 5.6 4.5 5.9 4.9 4.3 4.6 5.8 5.1 5.4 5.5 4.9 4.5 5.7 5.7 5.0
##    [409] 5.7 5.7 4.9 4.2 4.8 5.0 5.4 5.0 5.8 5.4 4.0 4.5 4.8 5.6 4.0 4.5 4.9
##    [426] 5.7 5.3 6.0 6.0 4.8 5.1 4.4 5.4 4.5 5.9 5.9 6.0 4.0 4.8 4.1 4.2 4.3
##    [443] 5.9 6.0 5.8 4.2 4.6 5.4 5.1 5.2 5.1 5.6 4.8 4.2 5.7 5.6 4.3 6.0 5.3
##    [460] 5.1 4.5 5.2 5.4 5.2 5.8 5.6 5.5 4.6 5.0 6.0 4.2 4.0 4.0 5.5 5.9 4.9
##    [477] 4.4 5.4 4.1 4.7 5.3 4.4 4.1 5.5 5.4 4.2 4.5 5.2 4.9 4.1 4.9 5.5 4.9
##    [494] 4.2 5.0 5.8 5.8 4.3 5.5 5.7 4.2 5.8 6.0 4.9 5.8 4.2 4.1 5.7 4.0 4.9
##    [511] 4.4 5.9 5.0 5.0 5.4 5.1 6.0 5.3 5.0 4.9 4.1 4.5 5.0 5.4 4.8 5.1 5.0
##    [528] 4.9 5.4 5.7 5.8 4.9 4.8 5.9 4.3 5.1 4.3 4.2 5.4 4.6 4.4 4.0 4.8 4.1
##    [545] 5.3 5.0 4.1 4.6 4.9 5.9 5.8 5.3 4.9 5.0 5.3 4.3 5.9 4.3 4.7 4.5 4.1
##    [562] 4.9 5.8 4.5 5.8 5.4 6.0 4.2 5.7 5.5 5.7 4.2 5.6 4.4 5.2 5.0 6.0 5.6
##    [579] 5.7 4.7 6.0 4.5 5.7 4.5 4.7 5.2 5.0 5.4 5.1 4.8 6.0 5.7 5.0 4.9 4.6
##    [596] 5.3 4.3 5.0 5.0 5.5 5.3 4.0 4.2 5.3 4.4 5.4 5.5 4.2 5.4 4.6 4.5 6.0
##    [613] 4.6 4.0 4.5 5.3 5.6 5.2 5.1 5.9 5.2 5.5 4.0 4.8 4.6 4.6 5.0 5.8 4.9
##    [630] 4.5 5.9 4.9 6.0 5.8 4.4 4.2 5.7 5.0 4.6 6.0 5.9 4.8 5.8 4.1 4.5 5.1
##    [647] 4.8 4.8 5.7 5.3 5.7 4.5 4.1 5.2 4.5 5.9 4.2 4.4 4.7 4.7 5.9 5.4 4.8
##    [664] 4.8 5.6 4.1 5.3 6.0 4.4 5.9 5.1 4.0 5.2 5.1 4.9 5.7 5.5 5.8 5.4 5.3
##    [681] 5.1 5.7 4.4 4.7 6.0 5.0 5.5 4.9 5.2 5.2 5.3 4.6 4.2 5.5 5.5 5.8 4.6
##    [698] 5.1 4.8 5.7 4.7 4.2 5.4 4.6 4.3 5.0 5.5 5.6 4.5 4.6 5.0 5.0 6.0 5.0
##    [715] 5.0 5.4 5.2 4.9 4.3 5.7 4.0 4.8 4.5 4.6 5.6 5.4 6.0 4.1 5.3 4.3 5.1
##    [732] 4.5 5.0 5.3 4.8 4.2 5.5 4.3 4.0 5.0 4.1 5.6 5.5 4.6 5.8 5.4 5.7 5.2
##    [749] 4.2 5.2 5.6 5.8 5.3 4.7 4.5 5.6 4.3 4.2 4.6 5.9 4.8 4.8 5.3 5.4 5.5
##    [766] 5.9 4.5 5.7 5.3 4.4 4.0 5.8 4.2 5.2 4.9 5.1 5.3 4.5 5.7 5.0 5.9 4.9
```

```
## [783] 5.2 5.7 4.1 4.6 4.5 4.3 5.7 6.0 5.4 4.7 4.7 5.3 4.9 6.0 5.6 4.6 5.7
## [800] 5.4 4.1 5.7 5.6 4.0 5.2 4.6 5.6 5.1 4.0 5.6 4.2 4.4 4.0 5.6 5.6 5.2
## [817] 5.0 4.0 4.2 4.3 5.9 5.0 4.6 6.0 5.0 5.1 4.6 4.8 5.8 4.8 4.0 4.5 4.4
## [834] 4.6 4.4 4.1 4.7 4.2 4.5 4.0 4.2 5.7 4.4 5.3 5.2 5.5 4.0 4.7 5.7 5.7
## [851] 5.3 4.7 4.9 5.9 5.4 5.1 4.6 4.8 4.6 4.6 5.8 5.6 4.5 5.0 6.0 5.4 4.9
## [868] 4.4 5.5 4.6 4.1 6.0 4.4 4.3 4.5 4.4 5.4 5.5 5.3 4.4 4.2 5.6 4.2 5.4
## [885] 4.4 4.8 4.0 5.5 5.7 4.9 4.7 4.4 6.0 4.8 5.6 5.8 4.0 5.7 4.1 5.9 5.8
## [902] 5.1 5.4 4.1 4.5 4.9 5.6 5.5 5.5 4.0 4.3 5.2 5.0 4.2 5.7 4.9 4.0 5.7
## [919] 5.8 4.2 5.8 4.3 4.9 5.9 6.0 4.8 4.7 4.9 6.0 5.4 5.1 5.7 5.1 5.5 5.0
## [936] 5.7 4.7 5.8 5.8 4.5 5.2 4.5 5.2 4.8 4.7 5.7 4.3 5.0 4.2 5.7 4.4 5.8
## [953] 5.4 4.4 6.0 5.2 4.7 4.2 4.2 4.1 4.5 5.4 4.4 5.3 4.9 4.2 5.8 5.3 5.5
## [970] 5.6 5.1 4.4 5.8 4.3 5.5 4.1 5.3 4.1 4.1 5.4 4.7 5.5 5.2 5.7 5.2 5.2
## [987] 5.3 4.2 4.2 5.3 5.7 4.7 5.1 4.9 4.9 5.2 5.2 4.7 5.7 4.1
```

Exercise 10

```
f_hat <- function(x) {-180.4243 + 46.2822*x}
y_hat <- f_hat(sim_dat)
y_hat
```

```
## [1] 78.75602 32.47382 9.33272 4.70450 18.58916 64.87136 4.70450
## [8] 97.26890 41.73026 18.58916 18.58916 60.24314 92.64068 69.49958
## [15] 97.26890 23.21738 27.84560 4.70450 60.24314 27.84560 64.87136
## [22] 50.98670 37.10204 69.49958 18.58916 37.10204 13.96094 50.98670
## [29] 9.33272 41.73026 55.61492 83.38424 60.24314 50.98670 55.61492
## [36] 27.84560 32.47382 18.58916 55.61492 50.98670 64.87136 92.64068
## [43] 9.33272 97.26890 50.98670 9.33272 55.61492 4.70450 32.47382
## [50] 88.01246 4.70450 69.49958 92.64068 60.24314 60.24314 23.21738
## [57] 4.70450 32.47382 46.35848 32.47382 27.84560 64.87136 41.73026
## [64] 27.84560 4.70450 27.84560 60.24314 50.98670 69.49958 23.21738
## [71] 9.33272 74.12780 60.24314 37.10204 74.12780 50.98670 32.47382
## [78] 74.12780 46.35848 27.84560 46.35848 32.47382 13.96094 27.84560
## [85] 50.98670 46.35848 88.01246 74.12780 69.49958 78.75602 83.38424
## [92] 78.75602 50.98670 60.24314 64.87136 83.38424 78.75602 88.01246
## [99] 37.10204 69.49958 88.01246 13.96094 64.87136 55.61492 13.96094
## [106] 23.21738 46.35848 50.98670 50.98670 69.49958 55.61492 9.33272
## [113] 23.21738 69.49958 60.24314 41.73026 64.87136 18.58916 37.10204
## [120] 41.73026 92.64068 55.61492 50.98670 37.10204 50.98670 74.12780
## [127] 60.24314 23.21738 41.73026 4.70450 60.24314 9.33272 18.58916
## [134] 9.33272 69.49958 4.70450 23.21738 88.01246 9.33272 23.21738
## [141] 27.84560 23.21738 69.49958 92.64068 55.61492 46.35848 18.58916
## [148] 88.01246 27.84560 18.58916 41.73026 55.61492 64.87136 4.70450
## [155] 92.64068 88.01246 41.73026 92.64068 97.26890 4.70450 41.73026
## [162] 18.58916 55.61492 92.64068 13.96094 55.61492 13.96094 4.70450
## [169] 41.73026 83.38424 78.75602 18.58916 64.87136 13.96094 4.70450
## [176] 13.96094 83.38424 60.24314 83.38424 78.75602 41.73026 88.01246
## [183] 50.98670 83.38424 64.87136 74.12780 13.96094 23.21738 55.61492
## [190] 92.64068 23.21738 88.01246 46.35848 97.26890 92.64068 88.01246
## [197] 64.87136 74.12780 60.24314 46.35848 13.96094 97.26890 27.84560
## [204] 27.84560 92.64068 55.61492 64.87136 64.87136 97.26890 41.73026
## [211] 69.49958 23.21738 60.24314 64.87136 55.61492 23.21738 46.35848
## [218] 9.33272 4.70450 92.64068 88.01246 32.47382 74.12780 60.24314
## [225] 23.21738 50.98670 92.64068 88.01246 4.70450 32.47382 55.61492
## [232] 97.26890 37.10204 18.58916 78.75602 18.58916 13.96094 64.87136
```

```

## [239] 92.64068 32.47382 55.61492 13.96094 13.96094 18.58916 46.35848
## [246] 37.10204 50.98670 9.33272 4.70450 74.12780 41.73026 46.35848
## [253] 92.64068 23.21738 50.98670 64.87136 27.84560 64.87136 60.24314
## [260] 9.33272 9.33272 41.73026 50.98670 83.38424 92.64068 60.24314
## [267] 64.87136 83.38424 92.64068 60.24314 13.96094 23.21738 13.96094
## [274] 4.70450 46.35848 37.10204 92.64068 32.47382 92.64068 46.35848
## [281] 50.98670 4.70450 78.75602 64.87136 23.21738 60.24314 64.87136
## [288] 50.98670 78.75602 92.64068 83.38424 83.38424 13.96094 13.96094
## [295] 92.64068 69.49958 27.84560 41.73026 60.24314 55.61492 23.21738
## [302] 55.61492 60.24314 46.35848 78.75602 92.64068 64.87136 9.33272
## [309] 88.01246 23.21738 13.96094 88.01246 13.96094 13.96094 18.58916
## [316] 9.33272 92.64068 46.35848 46.35848 9.33272 69.49958 37.10204
## [323] 13.96094 23.21738 4.70450 92.64068 83.38424 18.58916 69.49958
## [330] 83.38424 37.10204 4.70450 97.26890 83.38424 4.70450 74.12780
## [337] 78.75602 74.12780 18.58916 27.84560 46.35848 60.24314 23.21738
## [344] 83.38424 60.24314 74.12780 55.61492 46.35848 23.21738 78.75602
## [351] 83.38424 41.73026 69.49958 9.33272 23.21738 41.73026 92.64068
## [358] 27.84560 37.10204 23.21738 27.84560 46.35848 55.61492 83.38424
## [365] 4.70450 78.75602 27.84560 83.38424 64.87136 9.33272 74.12780
## [372] 74.12780 9.33272 69.49958 92.64068 41.73026 13.96094 9.33272
## [379] 41.73026 88.01246 69.49958 41.73026 64.87136 23.21738 69.49958
## [386] 27.84560 9.33272 64.87136 46.35848 32.47382 78.75602 18.58916
## [393] 64.87136 78.75602 27.84560 92.64068 46.35848 18.58916 32.47382
## [400] 88.01246 55.61492 69.49958 74.12780 46.35848 27.84560 83.38424
## [407] 83.38424 50.98670 83.38424 83.38424 46.35848 13.96094 41.73026
## [414] 50.98670 69.49958 50.98670 88.01246 69.49958 4.70450 27.84560
## [421] 41.73026 78.75602 4.70450 27.84560 46.35848 83.38424 64.87136
## [428] 97.26890 97.26890 41.73026 55.61492 23.21738 69.49958 27.84560
## [435] 92.64068 92.64068 97.26890 4.70450 41.73026 9.33272 13.96094
## [442] 18.58916 92.64068 97.26890 88.01246 13.96094 32.47382 69.49958
## [449] 55.61492 60.24314 55.61492 78.75602 41.73026 13.96094 83.38424
## [456] 78.75602 18.58916 97.26890 64.87136 55.61492 27.84560 60.24314
## [463] 69.49958 60.24314 88.01246 78.75602 74.12780 32.47382 50.98670
## [470] 97.26890 13.96094 4.70450 4.70450 74.12780 92.64068 46.35848
## [477] 23.21738 69.49958 9.33272 37.10204 64.87136 23.21738 9.33272
## [484] 74.12780 69.49958 13.96094 27.84560 60.24314 46.35848 9.33272
## [491] 46.35848 74.12780 46.35848 13.96094 50.98670 88.01246 88.01246
## [498] 18.58916 74.12780 83.38424 13.96094 88.01246 97.26890 46.35848
## [505] 88.01246 13.96094 9.33272 83.38424 4.70450 46.35848 23.21738
## [512] 92.64068 50.98670 50.98670 69.49958 55.61492 97.26890 64.87136
## [519] 50.98670 46.35848 9.33272 27.84560 50.98670 69.49958 41.73026
## [526] 55.61492 50.98670 46.35848 69.49958 83.38424 88.01246 46.35848
## [533] 41.73026 92.64068 18.58916 55.61492 18.58916 13.96094 69.49958
## [540] 32.47382 23.21738 4.70450 41.73026 9.33272 64.87136 50.98670
## [547] 9.33272 32.47382 46.35848 92.64068 88.01246 64.87136 46.35848
## [554] 50.98670 64.87136 18.58916 92.64068 18.58916 37.10204 27.84560
## [561] 9.33272 46.35848 88.01246 27.84560 88.01246 69.49958 97.26890
## [568] 13.96094 83.38424 74.12780 83.38424 13.96094 78.75602 23.21738
## [575] 60.24314 50.98670 97.26890 78.75602 83.38424 37.10204 97.26890
## [582] 27.84560 83.38424 27.84560 37.10204 60.24314 50.98670 69.49958
## [589] 55.61492 41.73026 97.26890 83.38424 50.98670 46.35848 32.47382
## [596] 64.87136 18.58916 50.98670 50.98670 74.12780 64.87136 4.70450
## [603] 13.96094 64.87136 23.21738 69.49958 74.12780 13.96094 69.49958
## [610] 32.47382 27.84560 97.26890 32.47382 4.70450 27.84560 64.87136

```

```

## [617] 78.75602 60.24314 55.61492 92.64068 60.24314 74.12780 4.70450
## [624] 41.73026 32.47382 32.47382 50.98670 88.01246 46.35848 27.84560
## [631] 92.64068 46.35848 97.26890 88.01246 23.21738 13.96094 83.38424
## [638] 50.98670 32.47382 97.26890 92.64068 41.73026 88.01246 9.33272
## [645] 27.84560 55.61492 41.73026 41.73026 83.38424 64.87136 83.38424
## [652] 27.84560 9.33272 60.24314 27.84560 92.64068 13.96094 23.21738
## [659] 37.10204 37.10204 92.64068 69.49958 41.73026 41.73026 78.75602
## [666] 9.33272 64.87136 97.26890 23.21738 92.64068 55.61492 4.70450
## [673] 60.24314 55.61492 46.35848 83.38424 74.12780 88.01246 69.49958
## [680] 64.87136 55.61492 83.38424 23.21738 37.10204 97.26890 50.98670
## [687] 74.12780 46.35848 60.24314 60.24314 64.87136 32.47382 13.96094
## [694] 74.12780 74.12780 88.01246 32.47382 55.61492 41.73026 83.38424
## [701] 37.10204 13.96094 69.49958 32.47382 18.58916 50.98670 74.12780
## [708] 78.75602 27.84560 32.47382 50.98670 50.98670 97.26890 50.98670
## [715] 50.98670 69.49958 60.24314 46.35848 18.58916 83.38424 4.70450
## [722] 41.73026 27.84560 32.47382 78.75602 69.49958 97.26890 9.33272
## [729] 64.87136 18.58916 55.61492 27.84560 50.98670 64.87136 41.73026
## [736] 13.96094 74.12780 18.58916 4.70450 50.98670 9.33272 78.75602
## [743] 74.12780 32.47382 88.01246 69.49958 83.38424 60.24314 13.96094
## [750] 60.24314 78.75602 88.01246 64.87136 37.10204 27.84560 78.75602
## [757] 18.58916 13.96094 32.47382 92.64068 41.73026 41.73026 64.87136
## [764] 69.49958 74.12780 92.64068 27.84560 83.38424 64.87136 23.21738
## [771] 4.70450 88.01246 13.96094 60.24314 46.35848 55.61492 64.87136
## [778] 27.84560 83.38424 50.98670 92.64068 46.35848 60.24314 83.38424
## [785] 9.33272 32.47382 27.84560 18.58916 83.38424 97.26890 69.49958
## [792] 37.10204 37.10204 64.87136 46.35848 97.26890 78.75602 32.47382
## [799] 83.38424 69.49958 9.33272 83.38424 78.75602 4.70450 60.24314
## [806] 32.47382 78.75602 55.61492 4.70450 78.75602 13.96094 23.21738
## [813] 4.70450 78.75602 78.75602 60.24314 50.98670 4.70450 13.96094
## [820] 18.58916 92.64068 50.98670 32.47382 97.26890 50.98670 55.61492
## [827] 32.47382 41.73026 88.01246 41.73026 4.70450 27.84560 23.21738
## [834] 32.47382 23.21738 9.33272 37.10204 13.96094 27.84560 4.70450
## [841] 13.96094 83.38424 23.21738 64.87136 60.24314 74.12780 4.70450
## [848] 37.10204 83.38424 83.38424 64.87136 37.10204 46.35848 92.64068
## [855] 69.49958 55.61492 32.47382 41.73026 32.47382 32.47382 88.01246
## [862] 78.75602 27.84560 50.98670 97.26890 69.49958 46.35848 23.21738
## [869] 74.12780 32.47382 9.33272 97.26890 23.21738 18.58916 27.84560
## [876] 23.21738 69.49958 74.12780 64.87136 23.21738 13.96094 78.75602
## [883] 13.96094 69.49958 23.21738 41.73026 4.70450 74.12780 83.38424
## [890] 46.35848 37.10204 23.21738 97.26890 41.73026 78.75602 88.01246
## [897] 4.70450 83.38424 9.33272 92.64068 88.01246 55.61492 69.49958
## [904] 9.33272 27.84560 46.35848 78.75602 74.12780 74.12780 4.70450
## [911] 18.58916 60.24314 50.98670 13.96094 83.38424 46.35848 4.70450
## [918] 83.38424 88.01246 13.96094 88.01246 18.58916 46.35848 92.64068
## [925] 97.26890 41.73026 37.10204 46.35848 97.26890 69.49958 55.61492
## [932] 83.38424 55.61492 74.12780 50.98670 83.38424 37.10204 88.01246
## [939] 88.01246 27.84560 60.24314 27.84560 60.24314 41.73026 37.10204
## [946] 83.38424 18.58916 50.98670 13.96094 83.38424 23.21738 88.01246
## [953] 69.49958 23.21738 97.26890 60.24314 37.10204 13.96094 13.96094
## [960] 9.33272 27.84560 69.49958 23.21738 64.87136 46.35848 13.96094
## [967] 88.01246 64.87136 74.12780 78.75602 55.61492 23.21738 88.01246
## [974] 18.58916 74.12780 9.33272 64.87136 9.33272 9.33272 69.49958
## [981] 37.10204 74.12780 60.24314 83.38424 60.24314 60.24314 64.87136
## [988] 13.96094 13.96094 64.87136 83.38424 37.10204 55.61492 46.35848

```



```
## [995] 46.35848 60.24314 60.24314 37.10204 83.38424 9.33272
```

Exercise 11

```
res <- m1$res  
y_pop <- res + y_hat  
y_pop
```

##	1	2	3	4	5	6
##	78.0257349	33.5128614	-17.1668915	14.3717625	24.8846436	72.1668436
##	7	8	9	10	11	12
##	5.9742149	89.0514993	39.6281960	18.9999803	8.3717593	43.7692971
##	13	14	15	16	17	18
##	88.4232793	56.2821793	89.3717410	15.6282003	13.5766620	-2.1411218
##	19	20	21	22	23	24
##	50.0257393	22.6281993	54.0257382	49.0257414	31.8846393	54.3975160
##	25	26	27	28	29	30
##	6.0895485	47.3975236	-0.5129029	69.7435306	6.4870982	35.5128593
##	31	32	33	34	35	36
##	44.1410771	76.2821760	60.5128549	56.2821836	37.7692982	22.2564203
##	37	38	39	40	41	42
##	21.6281982	3.1153171	63.2821825	38.7692993	61.7692960	92.1668371
##	43	44	45	46	47	48
##	9.1153193	104.6797203	45.5128571	-3.0257861	51.7692982	-0.5129007
##	49	50	51	52	53	54
##	34.8846403	105.3975117	3.7435414	79.7950636	86.1668371	54.6539603
##	55	56	57	58	59	60
##	62.2821814	30.9999793	-6.1411218	38.7693036	46.1410793	25.8846403
##	61	62	63	64	65	66
##	25.7435360	67.5386225	51.7435328	27.3717571	-1.6540061	15.7435360
##	67	68	69	70	71	72
##	63.9104025	44.9999728	68.6539582	40.0895474	15.6282036	76.2821782
##	73	74	75	76	77	78
##	55.6539603	51.8588706	85.9103993	47.3975203	34.1410825	69.2821782
##	79	80	81	82	83	84
##	47.3975214	22.9742096	51.1153106	31.6281982	19.4870971	24.2564203
##	85	86	87	88	89	90
##	57.2821836	42.7693003	82.9103960	72.2821782	71.5386214	102.1668403
##	91	92	93	94	95	96
##	88.7692917	63.6539560	54.7435306	60.2821814	71.9104014	96.6797236
##	97	98	99	100	101	102
##	71.9103982	81.7692906	26.4870917	65.3975160	90.6797225	6.4870971
##	103	104	105	106	107	108
##	57.7692960	40.5128560	10.4870971	24.2564214	48.1410793	40.5128571
##	109	110	111	112	113	114
##	43.6024253	53.5128528	55.7692982	9.3717614	34.5128636	66.7692949
##	115	116	117	118	119	120
##	58.0257393	52.7693014	66.9999696	-4.5129040	47.3717549	40.7693014
##	121	122	123	124	125	126
##	78.9103949	48.0257403	40.8846360	32.2564182	45.7692993	61.5128517
##	127	128	129	130	131	132
##	67.2821814	41.2306528	58.9999749	5.1153203	54.3975182	19.3717614
##	133	134	135	136	137	138
##	11.7435382	6.8588771	64.9104003	-20.3975640	35.6024317	69.5386171

##	139	140	141	142	143	144
##	-6.0257861	26.2564214	38.3717571	32.5128636	64.5128528	103.4232793
##	145	146	147	148	149	150
##	64.6539614	42.3975214	17.3717593	83.6539539	25.9742096	32.8846436
##	151	152	153	154	155	156
##	64.3459853	61.8330977	70.2821803	10.7435414	71.5386160	89.9103960
##	157	158	159	160	161	162
##	41.7693014	93.0515003	81.9103939	2.2306571	51.3975225	10.8588749
##	163	164	165	166	167	168
##	53.1410771	92.1668371	7.2306549	41.6281928	33.2048863	12.8331096
##	169	170	171	172	173	174
##	45.6281960	82.5386182	84.1668403	27.1153171	66.5386225	18.6282025
##	175	176	177	178	179	180
##	5.3717625	20.5766653	82.3975128	54.3975182	94.0515025	73.2821771
##	181	182	183	184	185	186
##	60.8846382	94.4232803	44.7692993	84.4232814	69.5386225	89.7692939
##	187	188	189	190	191	192
##	6.3717603	30.8588739	47.1410771	89.1668371	29.3459896	85.9103960
##	193	194	195	196	197	198
##	25.8846371	108.9361625	101.1668371	97.0515014	59.3975171	60.2821782
##	199	200	201	202	203	204
##	72.6539603	60.1153106	18.3717603	104.5643836	25.7435360	35.9999782
##	205	206	207	208	209	210
##	83.7950582	50.0257403	70.6281906	51.0257382	92.1668360	38.7693014
##	211	212	213	214	215	216
##	74.3975160	17.6282003	69.9104025	80.3717485	59.0257403	25.6282003
##	217	218	219	220	221	222
##	49.3975214	17.8588771	10.4870993	92.6797214	80.9103960	38.9999771
##	223	224	225	226	227	228
##	75.7692939	68.2821814	21.3717582	56.6281939	96.6797214	82.1668382
##	229	230	231	232	233	234
##	8.7177728	24.4870928	52.5128560	102.6797203	29.8846393	3.2306539
##	235	236	237	238	239	240
##	65.9103982	25.8846436	10.7435393	76.8846328	82.5386160	31.7435349
##	241	242	243	244	245	246
##	60.7692982	20.6282025	-34.9104504	28.8588749	47.3717528	32.6281971
##	247	248	249	250	251	252
##	58.3975203	-9.7693440	0.8331096	72.1668414	30.6281960	44.5128582
##	253	254	255	256	257	258
##	84.0257317	14.8588739	46.7692993	72.9104014	29.3717571	62.7692960
##	259	260	261	262	263	264
##	69.2821814	5.9742139	25.7177717	36.6281960	37.7692993	87.1668393
##	265	266	267	268	269	270
##	94.6797214	50.8846339	48.3975171	74.1668393	97.1668371	55.3975182
##	271	272	273	274	275	276
##	5.7435393	11.8588739	16.6282025	1.7435414	66.9742053	29.6281971
##	277	278	279	280	281	282
##	100.6539528	29.8846403	83.7950582	42.1153106	49.7692993	2.7435414
##	283	284	285	286	287	288
##	63.2821771	70.1668436	20.9999793	52.1410760	66.9104014	49.8846360
##	289	290	291	292	293	294
##	76.9103982	91.6539528	81.3975128	59.6539549	12.4870971	16.4870971
##	295	296	297	298	299	300
##	98.6539528	65.5128528	29.0895463	53.0257436	71.5386236	46.1410771

##	301	302	303	304	305	306
##	21.9999793	53.8846349	53.1410760	28.8846371	81.1668403	84.5386160
##	307	308	309	310	311	312
##	74.5386225	-17.0257861	91.5386171	7.4870949	35.6024339	89.3975117
##	313	314	315	316	317	318
##	26.4613285	7.1153182	26.2306539	0.1153193	96.3079425	34.4870896
##	319	320	321	322	323	324
##	43.5128582	-11.0257861	55.7692949	45.8588706	13.7435393	13.9999793
##	325	326	327	328	329	330
##	13.3459939	80.4232793	79.4232814	6.1153171	52.0257371	73.5128496
##	331	332	333	334	335	336
##	14.2306496	-12.1411218	85.0514993	60.3975128	7.0895517	60.2821782
##	337	338	339	340	341	342
##	76.9103982	72.6281885	6.7435382	7.7435360	46.3975214	31.5128549
##	343	344	345	346	347	348
##	8.3717582	76.7950603	55.6539603	66.5386203	34.1410771	30.5128582
##	349	350	351	352	353	354
##	1.2306528	56.2821771	65.9103971	22.3717539	77.1668425	3.2048874
##	355	356	357	358	359	360
##	7.1153160	39.7435328	86.0257317	35.7177674	25.6281971	5.1153160
##	361	362	363	364	365	366
##	27.8846414	63.6539636	53.1153085	83.7950603	-2.5129007	68.2821771
##	367	368	369	370	371	372
##	20.2306517	69.5386182	71.9104014	19.9999825	67.5128517	109.6281885
##	373	374	375	376	377	378
##	14.7177717	67.8846317	90.4232793	81.3459853	4.7435393	10.7177717
##	379	380	381	382	383	384
##	40.2564171	108.8846274	81.8846317	44.5128593	38.8846328	21.2306528
##	385	386	387	388	389	390
##	62.8846317	35.2306517	9.3717614	76.0257382	55.6539636	22.1153139
##	391	392	393	394	395	396
##	81.1668403	3.8588749	62.6539593	78.7950614	26.4870939	99.6797214
##	397	398	399	400	401	402
##	70.4870896	-2.3975672	61.0895453	116.1410696	42.5128560	73.0257371
##	403	404	405	406	407	408
##	72.5386203	41.8588685	34.6281993	80.7950603	87.7950603	41.8846360
##	409	410	411	412	413	414
##	86.1668393	90.4232814	53.5128582	29.2306549	36.9999749	47.3975203
##	415	416	417	418	419	420
##	61.2821793	46.3717517	81.7950593	86.2821793	9.1153203	5.1153149
##	421	422	423	424	425	426
##	50.1410803	79.9103982	17.3717625	30.2306517	53.2564160	78.9103971
##	427	428	429	430	431	432
##	71.7692960	76.7950571	89.0514993	50.7693014	56.2821825	8.1153160
##	433	434	435	436	437	438
##	81.7950636	19.1153149	88.4232793	88.0515003	95.0514993	-13.1411218
##	439	440	441	442	443	444
##	42.8846382	2.2306560	13.3717603	10.4870960	83.9103949	78.7950571
##	445	446	447	448	449	450
##	100.3975117	-4.1411240	34.8846403	104.8846317	85.4870874	57.3975182
##	451	452	453	454	455	456
##	48.0257403	81.1668403	27.6281960	13.3717603	84.7950603	66.2821771
##	457	458	459	460	461	462
##	13.7435382	84.4232782	82.3717485	62.1410771	15.6281993	109.2564128

##	463	464	465	466	467	468
##	81.2564106	52.0257393	83.3975117	72.5386193	71.9103993	15.7435349
##	469	470	471	472	473	474
##	49.7692993	99.1668360	7.8588760	10.4870993	-1.8846797	63.1410728
##	475	476	477	478	479	480
##	83.7950582	40.8846371	18.7177685	60.0257371	-5.5129018	26.8846393
##	481	482	483	484	485	486
##	55.0257382	23.8846425	17.6282036	66.7692939	73.1668425	32.7177706
##	487	488	489	490	491	492
##	45.6281993	61.9104025	42.9999739	-3.1411229	35.5128582	62.3975149
##	493	494	495	496	497	498
##	51.3975214	22.9999814	56.6539625	84.8846274	83.7950593	-3.1411251
##	499	500	501	502	503	504
##	83.1668414	77.6539549	13.6024339	74.5386171	103.4232782	47.1410793
##	505	506	507	508	509	510
##	83.5386171	18.9999814	0.2306560	111.1668393	27.8588782	47.9999739
##	511	512	513	514	515	516
##	20.1153160	110.5128474	43.8846360	50.5128571	74.1668425	46.5128560
##	517	518	519	520	521	522
##	85.7950571	71.2821803	47.7692993	53.2564160	15.7435403	40.5128625
##	523	524	525	526	527	528
##	43.8846360	57.6539582	48.6024274	44.7692982	62.5128571	60.7435317
##	529	530	531	532	533	534
##	74.2821793	81.2821760	86.8846274	52.8846371	53.0257436	84.1668371
##	535	536	537	538	539	540
##	10.8588749	54.5128560	14.4870960	19.9999814	86.9999685	33.9999771
##	541	542	543	544	545	546
##	35.0895474	5.7435414	35.8846382	2.4870982	43.1410749	36.5128571
##	547	548	549	550	551	552
##	19.8331085	33.9999771	25.3717528	91.0515003	93.4232803	61.1410749
##	553	554	555	556	557	558
##	32.2564160	49.8846360	50.1410749	21.6282014	94.6797214	43.9484431
##	559	560	561	562	563	564
##	32.6281971	24.9999782	11.9742139	26.2564160	66.6539539	40.9742096
##	565	566	567	568	569	570
##	95.0515014	78.5386214	87.4232782	20.7177706	78.5386182	74.3717463
##	571	572	573	574	575	576
##	110.1410706	7.2306549	79.9103982	0.2306528	63.0257393	41.1410782
##	577	578	579	580	581	582
##	106.7950571	74.9103982	97.1410706	35.4870917	97.5386149	27.2564203
##	583	584	585	586	587	588
##	82.7692917	9.7435360	24.6281971	69.5128549	41.5128571	71.2821793
##	589	590	591	592	593	594
##	50.1410771	35.1153117	98.6797203	79.1668393	43.8846360	51.6281949
##	595	596	597	598	599	600
##	38.1410825	65.3975171	7.2306539	63.2821836	45.8846360	69.0257360
##	601	602	603	604	605	606
##	93.6281906	-9.7693429	15.4870971	41.5128539	38.8331053	58.3975160
##	607	608	609	610	611	612
##	68.9103993	5.6024339	70.7692949	26.9999771	22.6281993	90.5386149
##	613	614	615	616	617	618
##	23.3717560	8.7435414	26.2306517	79.1410749	79.3975139	53.6281917
##	619	620	621	622	623	624
##	65.0257403	99.4232793	56.1410760	61.0257360	25.8331096	57.1153117

##	625	626	627	628	629	630
##	24.1153139	27.6281982	54.8846360	97.0515014	40.7435317	43.1153149
##	631	632	633	634	635	636
##	96.6797214	60.6281949	94.0514993	86.1668382	22.6282003	64.2048863
##	637	638	639	640	641	642
##	94.6797236	35.9999728	46.5128614	95.7950571	78.1668371	41.8846382
##	643	644	645	646	647	648
##	69.0257328	14.2306560	31.3717571	63.8846349	53.2564171	35.5128593
##	649	650	651	652	653	654
##	91.6281863	63.9104014	101.8846285	29.1153149	24.5766663	64.0257393
##	655	656	657	658	659	660
##	19.7435360	87.1668371	37.3459917	12.7435371	38.2564182	46.1410814
##	661	662	663	664	665	666
##	93.9103949	71.1410739	72.6024274	31.7435328	85.7950614	18.0895506
##	667	668	669	670	671	672
##	59.1410749	104.0514993	21.4870949	100.4232793	58.2821825	3.3459939
##	673	674	675	676	677	678
##	51.7692971	59.7692982	-0.5129104	81.6539549	82.6539571	83.2821749
##	679	680	681	682	683	684
##	49.3975160	68.5128539	50.7435296	87.1668393	23.7435371	35.5128603
##	685	686	687	688	689	690
##	91.0514993	64.0257414	78.7950625	47.3975214	73.2564128	60.9104025
##	691	692	693	694	695	696
##	66.2821803	39.2306506	9.9999814	82.2821782	79.9103993	80.7950593
##	697	698	699	700	701	702
##	46.4870928	61.9104036	24.9999749	83.3975128	44.1410814	44.7177706
##	703	704	705	706	707	708
##	77.2564106	14.6281982	14.9999803	42.1410782	52.6539571	94.1410717
##	709	710	711	712	713	714
##	40.6281993	29.8846403	50.8846360	81.2306463	96.9103939	17.3717517
##	715	716	717	718	719	720
##	31.5128571	64.2821793	67.5128549	49.3975214	38.4870960	85.0515025
##	721	722	723	724	725	726
##	0.6024360	54.0257436	24.4870939	31.4870928	66.9103982	69.2821793
##	727	728	729	730	731	732
##	106.5643836	11.4870982	50.7692960	15.3717593	61.5128560	22.7435360
##	733	734	735	736	737	738
##	60.2821836	64.9104014	31.8846382	20.8588760	63.2821782	24.2564225
##	739	740	741	742	743	744
##	-14.7693429	57.3975203	6.8588771	83.5128506	63.6539571	31.3717560
##	745	746	747	748	749	750
##	102.0257328	58.2564106	80.1668393	45.7692971	4.1153182	67.5386236
##	751	752	753	754	755	756
##	84.7950614	92.1410696	91.2306431	32.7435339	33.9999782	69.5386193
##	757	758	759	760	761	762
##	24.0895485	35.0895496	23.8588717	103.3079425	40.1410803	31.8846382
##	763	764	765	766	767	768
##	65.5386225	83.8846317	80.2564096	93.7950582	34.5128625	86.1668393
##	769	770	771	772	773	774
##	51.7692960	28.5128636	10.0895517	99.3079436	16.3717603	62.6539603
##	775	776	777	778	779	780
##	63.6539636	56.2821825	58.6539593	19.2564203	81.4232814	58.2821836
##	781	782	783	784	785	786
##	89.7950582	41.2564160	57.2564128	86.7950603	7.7177717	29.3717560

##	787	788	789	790	791	792
##	31.9742096	30.6024328	84.5386182	123.2821728	76.1668425	17.7435339
##	793	794	795	796	797	798
##	43.7693025	76.1668436	50.1410793	88.4232782	88.5386193	41.6281982
##	799	800	801	802	803	804
##	83.7950603	86.9104003	-5.6540072	92.5386182	63.9103982	11.4870993
##	805	806	807	808	809	810
##	67.3975182	45.1410825	110.3975139	45.5128560	4.1153203	78.2821771
##	811	812	813	814	815	816
##	24.4870971	11.9742106	-4.1411218	71.1668403	98.2821771	69.5386236
##	817	818	819	820	821	822
##	62.1410782	-7.0257851	11.4870971	20.2564225	96.2821739	53.8846360
##	823	824	825	826	827	828
##	46.1410825	98.6797203	44.7692993	62.9104036	5.7435349	43.5128593
##	829	830	831	832	833	834
##	98.7950593	30.2564171	18.7435414	33.5128625	27.2564214	39.7693036
##	835	836	837	838	839	840
##	28.8846425	16.9999825	28.9999760	16.2306549	16.2306517	-5.2822272
##	841	842	843	844	845	846
##	10.2306549	99.1668393	8.2306528	60.6281906	65.2821814	80.7950625
##	847	848	849	850	851	852
##	22.9742149	36.2564182	86.3975128	108.7692917	71.5386225	40.9999760
##	853	854	855	856	857	858
##	45.1153106	98.9103949	74.6539582	61.9104036	44.3717560	47.3975225
##	859	860	861	862	863	864
##	30.8846403	30.8846403	97.3079436	92.6539560	45.5128625	50.0257414
##	865	866	867	868	869	870
##	91.5386149	58.7692949	47.3975214	38.2564214	68.7435253	67.2048820
##	871	872	873	874	875	876
##	15.3717614	101.4232782	12.8588739	17.3717593	38.1410836	24.6282003
##	877	878	879	880	881	882
##	78.5386214	76.2821782	54.1410749	27.9999793	17.9999814	76.7950614
##	883	884	885	886	887	888
##	-6.0257872	70.5386214	26.9742106	43.8846382	7.2306571	45.1410728
##	889	890	891	892	893	894
##	84.3975128	47.8588685	23.3717549	30.2564214	93.1410674	48.1410803
##	895	896	897	898	899	900
##	85.4232825	86.7950593	32.3459939	100.5386182	18.9742139	120.9361636
##	901	902	903	904	905	906
##	87.1668382	44.6281928	70.1410739	12.4870982	29.8846414	69.3975214
##	907	908	909	910	911	912
##	98.7950614	43.8846306	76.5386203	-7.9104483	36.1153171	41.7692971
##	913	914	915	916	917	918
##	63.7692993	22.7435393	101.0257339	30.7435317	18.7435414	76.2821760
##	919	920	921	922	923	924
##	97.3079436	50.2048863	104.1410696	49.6024328	54.7693003	94.3079425
##	925	926	927	928	929	930
##	108.6539517	49.8846382	12.7435339	37.8588685	100.1668360	56.7692949
##	931	932	933	934	935	936
##	55.7692982	92.4232814	73.1410771	69.6539571	93.2306463	123.8846285
##	937	938	939	940	941	942
##	52.3975236	99.7692906	86.0515014	47.7435360	56.6539603	13.1153149
##	943	944	945	946	947	948
##	57.7692971	60.2306485	22.3717549	72.9103971	26.6282014	43.8588674

```
##      949      950      951      952      953      954
## -7.1411240 81.2821760 27.3717582 91.8846274 75.6539582 34.8846425
##      955      956      957      958      959      960
## 115.5643836 73.6539603 26.9999760 17.6024339 17.1153182 27.2306560
##      961      962      963      964      965      966
## 21.9999782 69.7692949 24.6282003 78.5386225 37.8588685 30.6282025
##      967      968      969      970      971      972
## 86.2821749 70.9104014 57.3975149 54.2564085 46.3975193 36.9742106
##      973      974      975      976      977      978
## 73.9103960 7.2306539 54.7692939 27.3717614 88.6539593 3.1153193
##      979      980      981      982      983      984
## 1.7435403 43.1410739 53.1153128 58.0257360 62.8846339 86.7950603
##      985      986      987      988      989      990
## 49.3975182 62.2821814 55.6281906 -0.7693451 21.2564236 60.7692960
##      991      992      993      994      995      996
## 82.7950603 45.5128603 74.2564139 58.6539636 42.3975214 59.0257393
##      997      998      999      1000
## 58.1410760 43.2564182 69.5386182 31.0637820
```

```
res_squared = (m1$res)^2
sum(res_squared)
```

```
## [1] 131999.6
```

```
sum(res_squared)/998
```

```
## [1] 132.2641
```

$$\hat{\sigma}^2 = RSS/n - 2$$

=

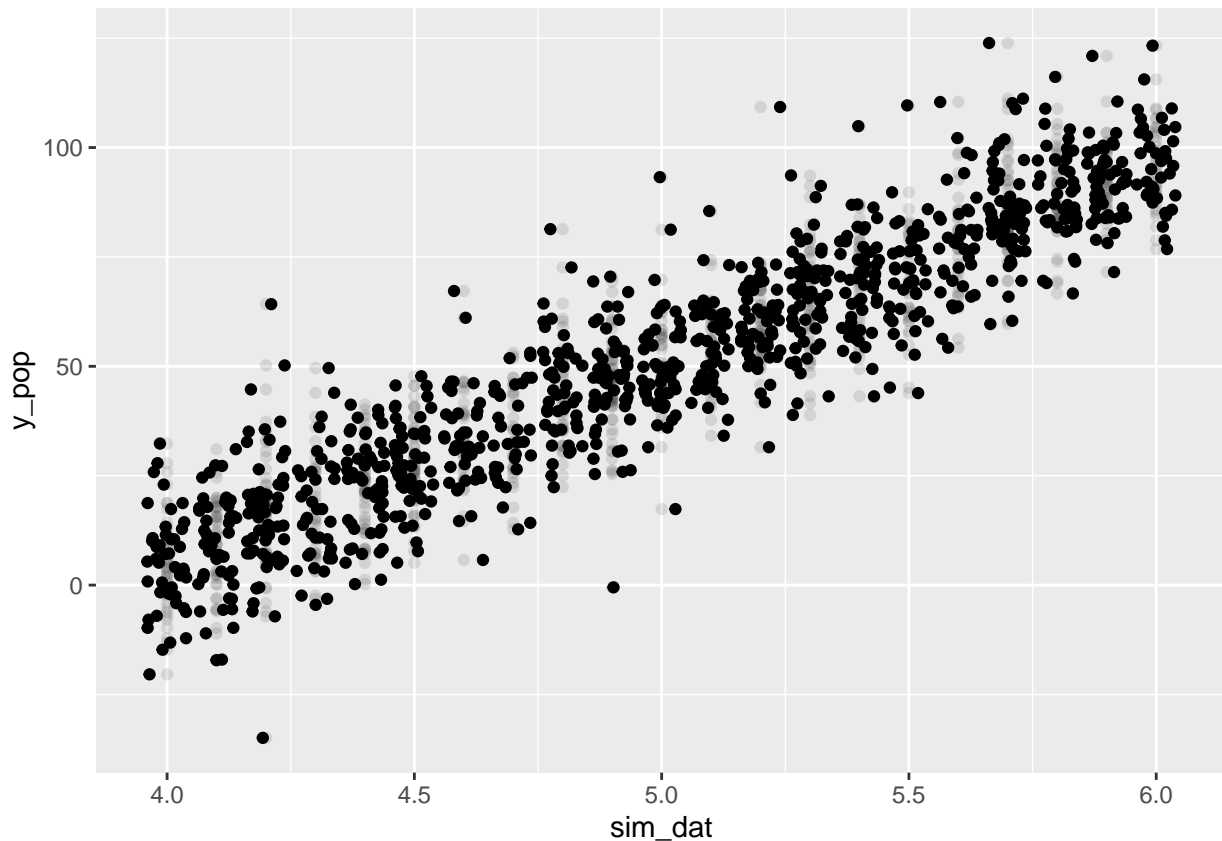
$$\hat{\sigma}^2 = 131999.6/998$$

= =

$$\hat{\sigma}^2 = 132.2641$$

Exercise 12

```
ggplot(quakes, aes(sim_dat, y_pop)) + geom_point(alpha = 0.1) + geom_jitter()
```



```
m2 = lm(y_pop ~ sim_dat, data = quakes)
summary(m2)
```

```
##
## Call:
## lm(formula = y_pop ~ sim_dat, data = quakes)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -49.032  -7.101  -0.388   6.754  50.084
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -179.4188     3.0548  -58.73  <2e-16 ***
## sim_dat      46.0809     0.6071   75.90  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 11.5 on 998 degrees of freedom
## Multiple R-squared:  0.8523, Adjusted R-squared:  0.8522
## F-statistic: 5761 on 1 and 998 DF, p-value: < 2.2e-16
```

As with the original data, there is a strong, positive, and linear relationship between the magnitude of an earthquake and number of stations detecting it in the simulated data. However, the intercept is higher, and the slope is a little bit lower, suggesting that for every unit increase in the magnitude of an earthquake on the Richter scale, the number of stations that detect it increases by 45.6. To increase my model, I might find an intercept that lies around the mean of -180.4243 and -177.0066 and a slope that lies between mean of 45.6 (the slope of the simulated data) and the original slope of 46.2822. ##### PROBLEM SET 2: ##### Exercise 1

1)

$$H_0$$

: While holding all other predictors constant, the amount of spending on TV advertising has no effect on sales.

The model suggests that the relationship between the amount of spending on TV advertising and sales is statistically significant, allowing us to reject the null hypothesis ($p < 0.0001$). Spending an additional \$1000 on TV advertising leads to an increase in sales by approximately 46 units.

2)

$$H_0$$

: While holding all other predictors constant, the amount of spending on radio advertising has no effect on sales.

The model also suggests that relationship between the amount of spending on radio advertising and sales is statistically significant, allowing us to reject the null hypothesis ($p < 0.0001$). Spending an additional \$1000 on radio advertising leads to an increases in sales by approximately 189 units.

3)

$$H_0$$

: While holding all other predictors constant, the amount of spending on newspaper advertising has no effect on sales.

While the model reports that spending an additional \$1000 on newspaper advertising may lead to a decrease in sales by one unit, this relationship is not statistically significant ($p = 0.8599$).

When there is no spending on either TV, radio, or newspaper advertising the model makes the statistically significant prediction that the mean value of sales would be 2.939 units ($p < 0.0001$).

Exercise 4

- Despite the fact that the true relationship is linear, I would expect the training RSS of the cubic regression to be lower because its greater flexibility gives it the capacity to fit greater variance in the training data.
- In this case, the greater flexibility cubic regression model will overfit the training data whereas the greater bias of the linear regression model is mediated by the true linear relationship. Consequently I would expect the test RSS of the linear regression to be lower.
- As in a), I would expect the training RSS of the cubic regression to be lower because its greater flexibility allows it to encompass the greater variance in training data regardless of the true relationship between X and Y.
- There is not enough information to tell because we do not know the linearity of the true relationship between X and Y. If the true relationship is closer to linear, the linear regression model would be a better fit with a lower test RSS. However, if the relationship is extremely non-linear then it is possible that the greater flexibility of the cubic regression model will be a better fit and yield a lower RSS.

Exercise 5

- If we are given the form of the i th fitted value as:

$$\hat{y} = x_i \hat{B}$$

where:

$$\hat{B} = (\sum_{i=1}^n x_i y_i) / (\sum_{i'=1}^n x_{i'}^2)$$

- Then, after substitution:

$$\hat{y} = (x_i \sum_{i'=1}^n x_{i'} y_{i'}) / (\sum_{i''=1}^n x_{i''}^2)$$

3. We were also given that:

$$\hat{y} = \sum_{i'=1}^n a_{i'} y_{i'}$$

,

So:

$$\hat{y} = \sum_{i'=1}^n a_{i'} y_{i'} = (x_i \sum_{i'=1}^n x_{i'} y_{i'}) / (\sum_{i''=1}^n x_{i''}^2)$$

=

$$a_{i'} y_{i'} = x_i x_{i'} y_{i'} / \sum_{i''=1}^n x_{i''}^2$$

=

$$a_i = x_i x_{i'} / \sum_{i''=1}^n x_{i''}^2$$

ADDITIONAL EXERCISES

Exercise 1

1) Variance:

a) We know that k-nearest regression is defined as:

$$[\hat{f}(x) = \frac{1}{k} \sum_{x_i \in \mathcal{N}(x)} y_i]$$

b) To find the variance:

$$Var(\hat{f}(x)) = Var\left(\frac{1}{k} \sum_{x_i \in \mathcal{N}(x)} y_i\right)$$

=

$$Var(\hat{f}(x)) = \frac{1}{k^2} Var\left(\sum_{x_i \in \mathcal{N}(x)} y_i\right)$$

=

$$Var(\hat{f}(x)) = \frac{1}{k^2} \sum_{x_i \in \mathcal{N}(x)} Var(y_i)$$

=

$$Var(\hat{f}(x)) = \frac{1}{k^2} \sum_{x_i \in \mathcal{N}(x)} (\sigma^2)$$

=

$$Var(\hat{f}(x)) = \frac{1}{k^2} k (\sigma^2)$$

=

$$Var(\hat{f}(x)) = \frac{k}{k^2} (\sigma^2)$$

=

$$Var(\hat{f}(x)) = \frac{\sigma^2}{k}$$

2) Bias

a) We know that k-nearest regression is defined as:

$$[\hat{f}(x) = \frac{1}{k} \sum_{x_i \in \mathcal{N}(x)} y_i]$$

b) To find bias:

$$\begin{aligned} Bias(\hat{f}(x)) &= E[f(x)] - E[\hat{f}(x)] \\ &= \\ Bias(\hat{f}(x)) &= E[f(x)] - E\left[\frac{1}{k} \sum_{x_i \in \mathcal{N}(x)} y_i\right] \\ &= \\ Bias(\hat{f}(x))^2 &= (E[f(x)] - E\left[\frac{1}{k} \sum_{x_i \in \mathcal{N}(x)} y_i\right])^2 \\ &= \\ Bias(\hat{f}(x))^2 &= (f(x) - (\frac{1}{k} \sum_{x_i \in \mathcal{N}(x)} y_i))^2 \end{aligned}$$

c) Decomposition of MSE:

$$\begin{aligned} MSE &= Var(f(x)) + Bias(\hat{f}(x))^2 + Var(\hat{f}x) \\ &= \\ MSE &= Var(f(x)) + (f(x) - (\frac{1}{k} \sum_{x_i \in \mathcal{N}(x)} y_i))^2 + \frac{\sigma^2}{k} \\ &= \\ MSE &= \sigma^2 + (f(x) - (\frac{1}{k} \sum_{x_i \in \mathcal{N}(x)} y_i))^2 + \frac{\sigma^2}{k} \end{aligned}$$

Exercise 2

```
library(tidyverse)
x <- c(1:3, 5:12)
y <- c(-7.1, -7.1, .5, -3.6, -2, -1.7, -4, -.2, -1.2, -1.2, -3.5)
y_mean = mean(y)
std_dev <- function(y) {(sum(y-y_mean)^2)/11}
std_dev <- std_dev(y)

my_fun <- function(k, x, y) {
  f_k <- rep(NA, length(k))
  for (i in 1:length(k)) {
    f_k[i] <- std_dev/k[i]
  }
  f_k
}

y_sum <- sum(y)
my_fun_2 <- function(k, x, y) {
  m_k <- rep(NA, length(k))
  for (i in 1:length(k)) {
```

```

    m_k[i] <- std_dev + (std_dev/k[i]) + (y_sum/k[i])^2
  }
  m_k
}

my_fun_3 <- function(k, x, y) {
  b_k <- rep(NA, length(k))
  for (i in 1:length(k)) {
    b_k[i] <- (y_sum/k[i])^2
  }
  b_k
}

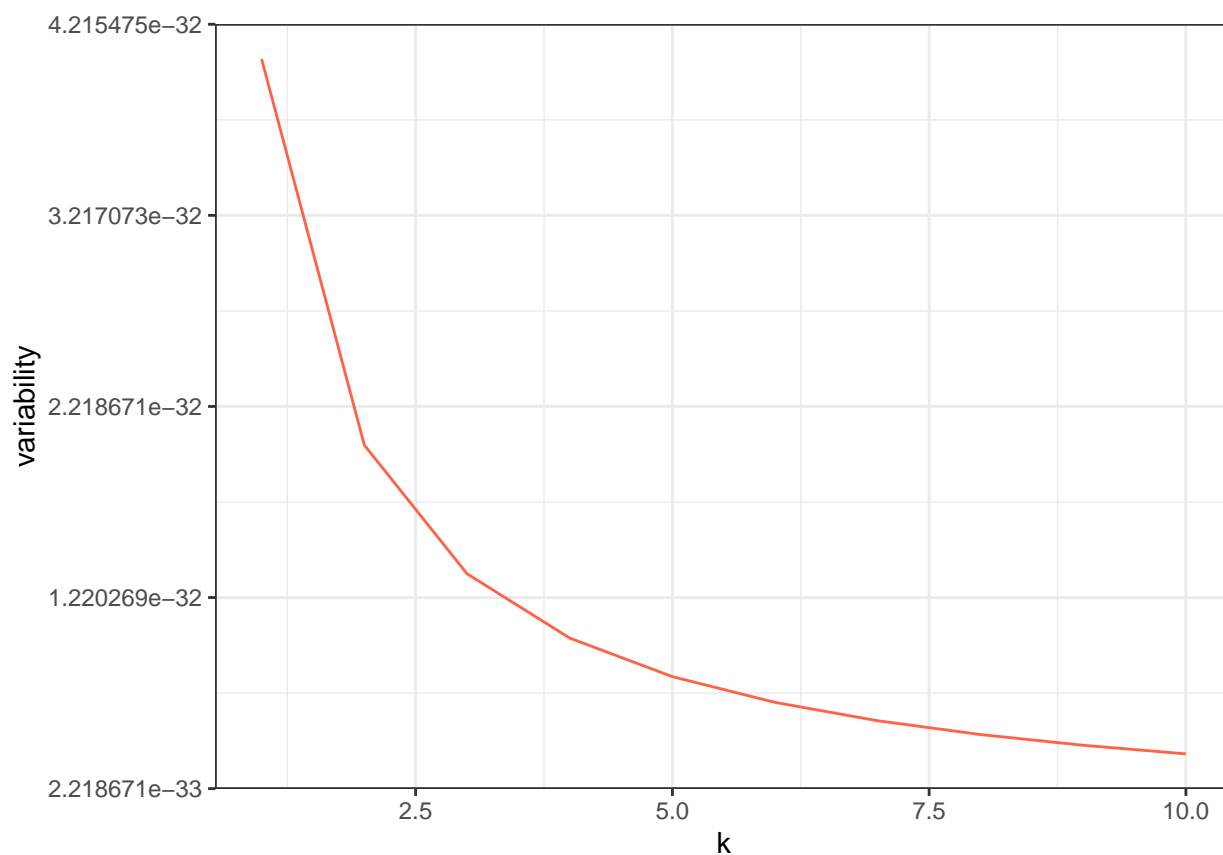
k <- 1:10
f_k <- my_fun(k, x, y)

m_k <- my_fun_2(k,x,y)

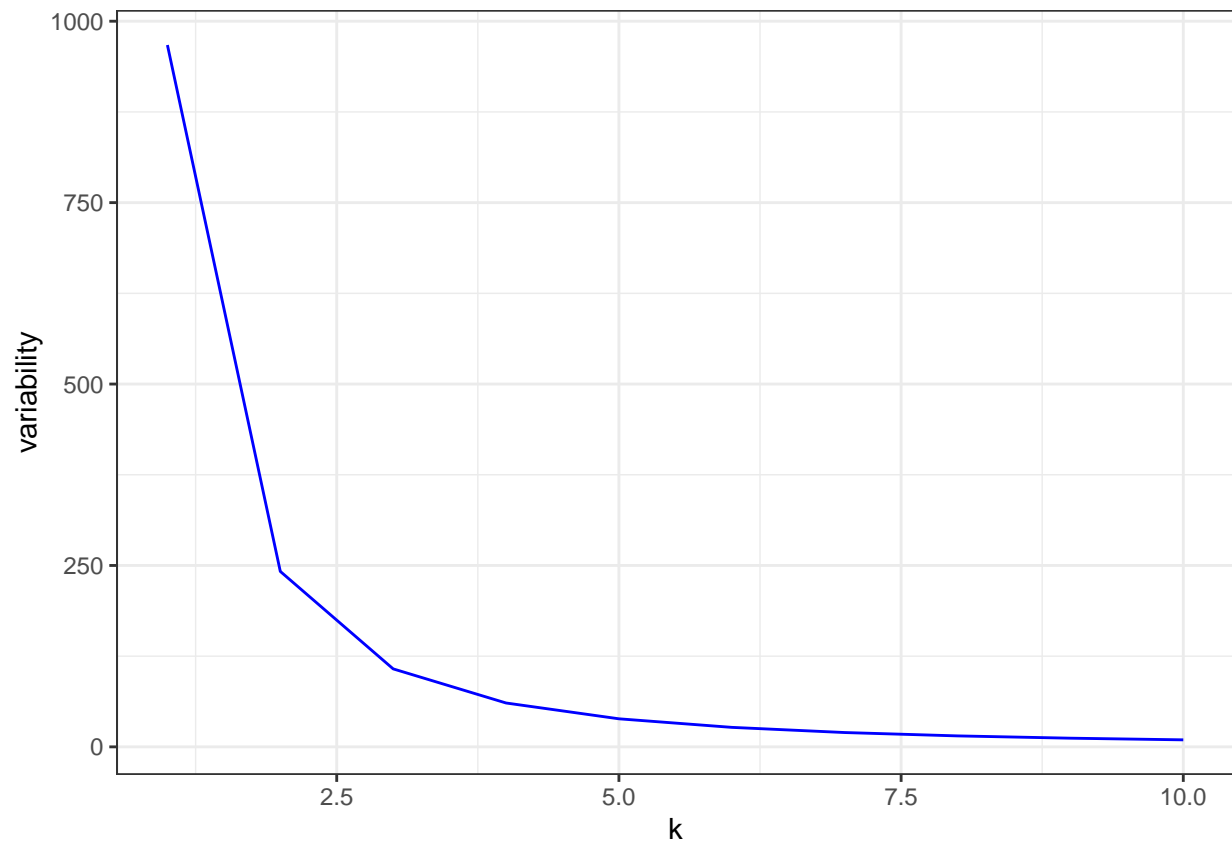
k <- 1:10
b_k <- my_fun_3(k, x, y)

df <- tibble(k = k, f_k = f_k)
ggplot(df, aes(x = k, y = f_k)) + geom_line (col = "tomato") +
  theme_bw() +
  ylab("variability")

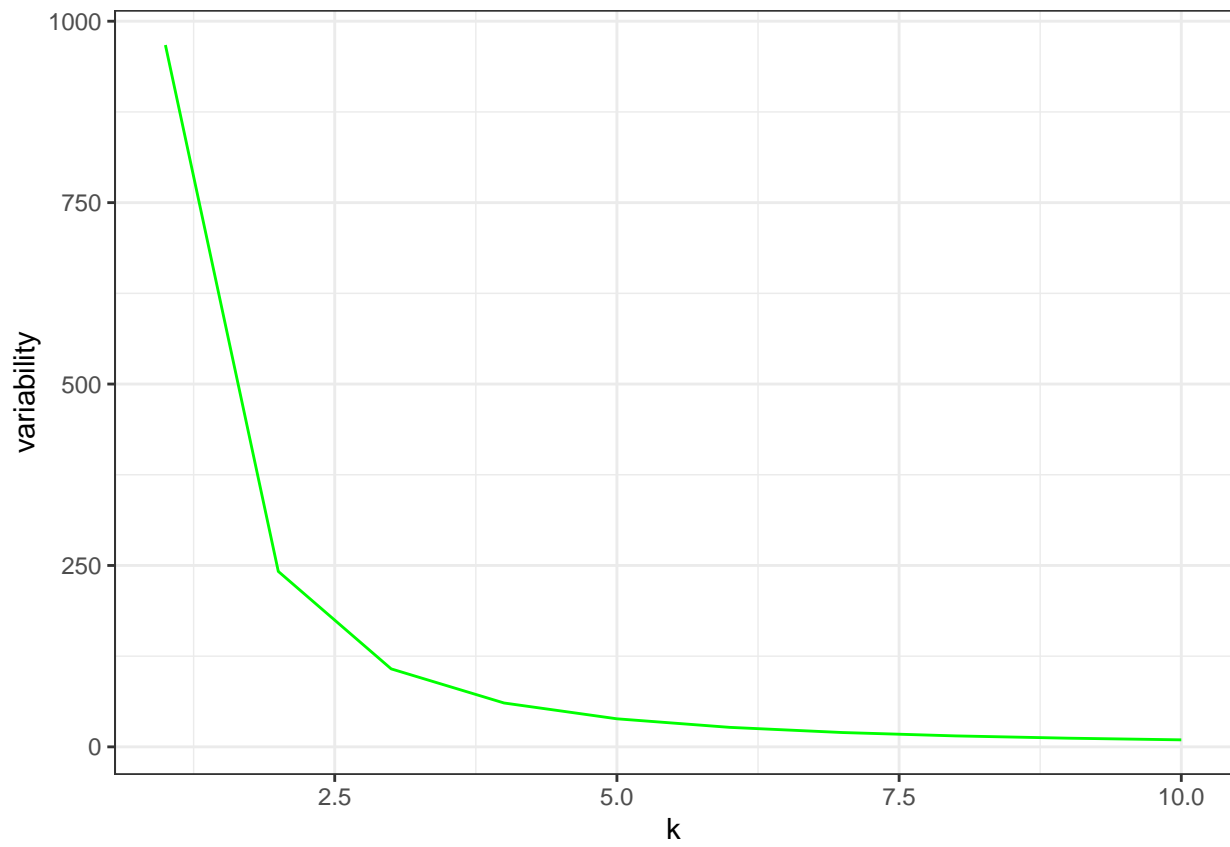
```



```
df <- tibble(k = k, m_k = m_k)
ggplot(df, aes(x = k, y = m_k)) + geom_line (col = "blue") +
  theme_bw() +
  ylab("variability")
```



```
df <- tibble(k = k, b_k = b_k)
ggplot(df, aes(x = k, y = b_k)) + geom_line (col = "green") +
  theme_bw() +
  ylab("variability")
```



Exercise 7

Exercise 7

Exercise 7

Exercise 7

Exercise 7

Exercise 7

Exercise 7