# Linear Models Assignment 7

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# Problem 1.

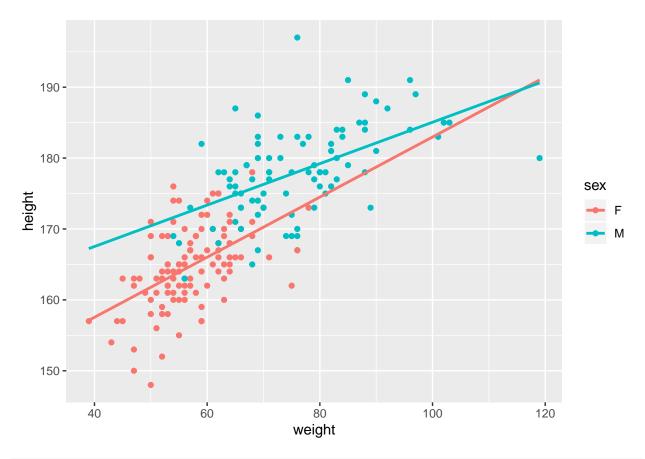
1.

### Test for Association

lm\_davis1 <- lm(weight ~ sex, data = davis)</pre>

```
summary(lm_davis1)
##
## Call:
## lm(formula = weight ~ sex, data = davis)
## Residuals:
      Min
               1Q Median
                               ЗQ
                                      Max
## -21.898 -6.395 -0.892
                            5.108 43.102
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 56.8919
                           0.8952
                                    63.55
                                            <2e-16 ***
## sexM
               19.0058
                           1.3461
                                    14.12
                                            <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 9.431 on 197 degrees of freedom
## Multiple R-squared: 0.5029, Adjusted R-squared: 0.5004
## F-statistic: 199.3 on 1 and 197 DF, p-value: < 2.2e-16
```

#### Test for Interaction



```
lm_davis2 <- lm(height ~ weight + sex + weight:sex, data = davis)
summary(lm_davis2)$coefficients</pre>
```

```
## Estimate Std. Error t value Pr(>|t|)
## (Intercept) 140.5913631 4.08463197 34.419591 8.673770e-85
## weight 0.4238098 0.07128015 5.945691 1.250947e-08
## sexM 15.2487377 5.42339531 2.811659 5.432264e-03
## weight:sexM -0.1316895 0.08507779 -1.547872 1.232745e-01
```

#### Comments

The association between weight and sex is statistically significant according to our F-statistic and p-value for the overall model significance. The interaction between weight and sex in predicting height seems possible visitually, but to further study if there is significant interaction we can look at the interaction term's t-statistic and corresponding p-value. After doing so, we receive an t-statistic of -1.548 and p-value of .1233, showing that the interaction is not statistically significant. These two differences show that association between two variables does not infer interaction between the same two in predicting a response variable.

#### 2.

I couldn't find a dataset where there was no significant association but a significant interaction between two independent variables (after looking for several hours), so instead I attempted to make a theoretical example. My apologies.

```
# theoretical setup
x_1 <- runif(1500, 1, 50)
x_2 <- rbinom(1500, 1, .1)
x_2 <- as.logical(x_2)
eps <- rnorm(1500, 0, 150)
y <- 500 + (3 * x_1) + (59*x_2) + (25 * x_1 * x_2) + eps
df_theor <- data.frame(y, x_1, x_2)</pre>
```

#### Test for Association

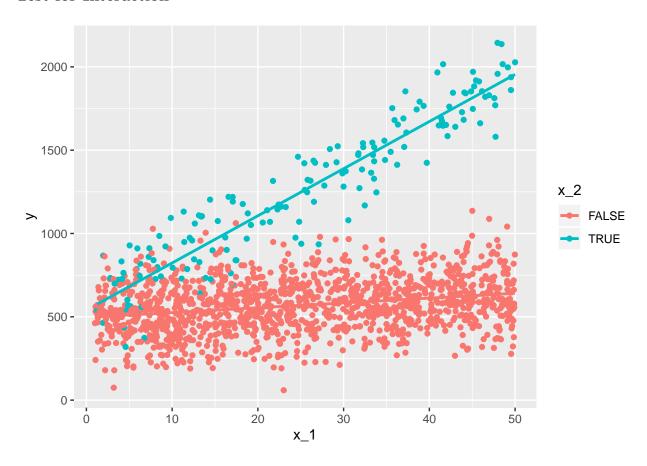
```
lm_theor1 <- lm(x_1 ~ x_2, data = df_theor)
summary(lm_theor1)

##
## Call:
## lm(formula = x_1 ~ x_2, data = df_theor)</pre>
```

```
##
## Residuals:
##
       Min
                 1Q Median
                                  3Q
## -24.1605 -12.9100 -0.8012 11.8314 25.3153
##
## Coefficients:
             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 24.6568 0.3873 63.670 <2e-16 ***
## x_2TRUE
              0.7748
                       1.1857 0.653
                                           0.514
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 14.18 on 1498 degrees of freedom
## Multiple R-squared: 0.0002849, Adjusted R-squared: -0.0003824
```

## F-statistic: 0.427 on 1 and 1498 DF, p-value: 0.5136

### Test for Interaction



```
lm_theor2 <- lm(y ~ x_1 + x_2 + x_1:x_2)
summary(lm_theor2)</pre>
```

```
##
## Call:
## lm(formula = y ~ x_1 + x_2 + x_1:x_2)
##
## Residuals:
##
      Min
               1Q Median
                               ЗQ
                                      Max
                            98.51 514.97
## -503.73 -98.93
                   -0.04
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 496.3980
                           8.3206 59.659
                                            <2e-16 ***
                           0.2929
                                            <2e-16 ***
## x_1
                2.8932
                                    9.877
## x_2TRUE
               43.8658
                          25.3354
                                    1.731
                                            0.0836 .
## x_1:x_2TRUE 25.3980
                           0.8661 29.324
                                            <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 151.2 on 1496 degrees of freedom
## Multiple R-squared: 0.7417, Adjusted R-squared: 0.7411
## F-statistic: 1432 on 3 and 1496 DF, p-value: < 2.2e-16
```

#### Comments

Clearly (by setup) there is no relationship between  $x_1$  and  $x_2$  (unless by random chance). However, there is a clear relationship between y and  $x_1$ ,  $x_2$ , and the interaction between  $x_1$  and  $x_2$ . We checked the association between  $x_1$  and  $x_2$  above, where we recieved an f-statistic of 2.594 and p-value of .1075. This association is statistically insignificant. However, when looking at the interaction between  $x_1$  and  $x_2$  in predicting y, we see that there is a significant interaction. We recieve a t-value of 26.949 and p-value of basically 0. Here we observe that interaction does not infer association.

# Problem 2.

1.

a.

$$\begin{bmatrix} 1 & 2 & 0 \\ 1 & 1 & 1 \\ 1 & 1 & 2 \\ 1 & 0 & 1 \end{bmatrix}' = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 2 & 1 & 1 & 0 \\ 0 & 1 & 2 & 1 \end{bmatrix}$$

b.

$$\left[\begin{array}{ccc} 1 & 3 & 1 \\ 2 & 2 & 5 \end{array}\right] \times \left[\begin{array}{ccc} 0 & 2 & 0 \\ 1 & 1 & 1 \\ 3 & 1 & 2 \end{array}\right] = \left[\begin{array}{ccc} 6 & 6 & 5 \\ 17 & 11 & 12 \end{array}\right]$$

c.

$$\begin{bmatrix} 1 & 2 & 0 \\ 1 & 1 & 1 \\ 1 & 1 & 2 \\ 1 & 0 & 1 \end{bmatrix} \times \begin{bmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \end{bmatrix} + \begin{bmatrix} \epsilon_1 \\ \epsilon_2 \\ \epsilon_3 \\ \epsilon_4 \end{bmatrix} = \begin{bmatrix} \beta_0 + 2\beta_1 + \epsilon_1 \\ \beta_0 + \beta_1 + \beta_2 + \epsilon_2 \\ \beta_0 + \beta_1 + 2\beta_2 + \epsilon_3 \\ \beta_0 + \beta_2 + \epsilon_4 \end{bmatrix}$$

**2**.

a.

```
A <- matrix(c(1, 4, 2, 5, 3, 6), 2, 3)
B <- matrix(c(-5, 3, 1, 0, 2, -4), 2, 3)

# Matrix Addition
A + B
```

```
## [,1] [,2] [,3]
## [1,] -4 3 5
## [2,] 7 5 2
```

```
# Matrix Subtraction
```

A - B

```
## [,1] [,2] [,3]
## [1,] 6 1 1
## [2,] 1 5 10
# Multiplication by scalar -1
## [,1] [,2] [,3]
## [1,] 5 -1 -2
## [2,] -3 0 4
b.
a \leftarrow t(c(2, 0, 1, 3))
b \leftarrow c(-1, 6, 0, 9)
a %*% b
## [,1]
## [1,] 25
c.
# p.1
A \leftarrow matrix(c(1, 3, 2, 4), 2, 2)
B \leftarrow matrix(c(0, 2, 3, 1), 2, 2)
## [,1] [,2]
## [1,] 4 5
## [2,] 8 13
# p.2
A \leftarrow matrix(c(2, 0, 0, 3), 2, 2)
B \leftarrow matrix(c((1/2), 0, 0, (1/3)), 2, 2)
A%*%B
## [,1] [,2]
## [1,] 1 0
## [2,] 0 1
\mathbf{d}.
A \leftarrow matrix(c(1, 2, 3, 4), 2, 2)
B \leftarrow matrix(c(5, 6, 7, 8), 2, 2)
```

```
## [,1] [,2]
## [1,] 1 3
## [2,] 2 4
## [,1] [,2]
## [1,] 5 7
## [2,] 6 8
# results of the matrix multiplications below are not equal
A%*%B
## [,1] [,2]
## [1,] 23 31
## [2,] 34 46
B%*%A
## [,1] [,2]
## [1,] 19 43
## [2,] 22 50
e.
A_trans <- t(A%*%B)
B_trans <- t(B)%*%t(A)</pre>
# two matrices below are equal, verifying that (AB)' = B'A'
A_trans
## [,1] [,2]
## [1,] 23 34
## [2,] 31 46
B_trans
## [,1] [,2]
## [1,] 23 34
## [2,] 31 46
f.
A \leftarrow matrix(c(0, 2, 3, 1), 2, 2)
A_square <- A%*%A
A_cube <- A%*%A_square
A_square
```

```
[,1] [,2]
##
## [1,]
           6 3
## [2,]
A_cube
##
        [,1] [,2]
## [1,]
          6
## [2,]
          14
                13
\mathbf{g}.
A \leftarrow matrix(c(2, 1, 5, 3), 2, 2)
# inverse of A
A_inverse <- solve(A)
A_{inverse}
##
     [,1] [,2]
## [1,] 3 -5
## [2,]
                 2
          -1
b < -c(4, 5)
x <- A_inverse%*%b
##
       [,1]
## [1,] -13
## [2,] 6
h.
A \leftarrow matrix(c(1, 3, 2, 5), 2, 2)
b < -c(7, 2)
x \leftarrow solve(A)%*%b
Х
##
       [,1]
## [1,] -31
## [2,] 19
i.
A \leftarrow matrix(c(1, 3, 2, 6), 2, 2)
b < -c(7, 2)
#x <- solve(A)%*%b
```

R gives an error, stating that the system is singular. This system of equations has no solution.

## Problem 3.

1.

## Matrix transpose function

```
my.transpose <- function(A){
  dims <- dim(A)
  A_trans <- c()
  for (i in 1:dims[1]){
    vect <- A[i,]
    A_trans <- c(A_trans, vect)
  }
  A_trans <- matrix(A_trans, dims[2], dims[1])
  return(A_trans)
}</pre>
```

### Matrix multiplication function

```
my.product <- function(A, B){
    dims_A <- dim(A)
    dims_B <- dim(B)
    vect <- c()
    for (i in 1:dims_B[2]){
        for (j in 1:dims_A[1]){
            dot_prod <- sum(A[j,] * B[,i])
            vect <- c(vect, dot_prod)
        }
    }
    AB <- matrix(vect, dims_A[1], dims_B[2])
    return(AB)
}</pre>
```

# Functionality checks

```
# checking my.transpose()
A \leftarrow matrix(c(1, 1, 1, 1, 2, 1, 1, 0, 0, 1, 2, 1), 4, 3)
# my function output
my.transpose(A)
##
       [,1] [,2] [,3] [,4]
## [1,]
       1 1 1 1
## [2,]
       2 1
                  1
            1
## [3,]
       0
                   2
# t() output
t(A)
```

```
## [,1] [,2] [,3] [,4]
## [1,] 1 1 1 1
## [2,]
       2
                  1
## [3,]
         0
                   2
                       1
              1
# checking my.product()
B \leftarrow matrix(c(1, 2, 3, 2, 1, 5), 2, 3)
C \leftarrow matrix(c(0, 1, 3, 2, 1, 1, 0, 1, 2), 3, 3)
# my function output
my.product(B,C)
      [,1] [,2] [,3]
## [1,] 6 6 5
## [2,] 17 11
                  12
# %*% operator output
B%*%C
## [,1] [,2] [,3]
## [1,] 6 6 5
## [2,] 17 11 12
2.
```

```
my.is.symm <- function(A){
  transpose <- my.transpose(A)
  dims <- dim(A)
  n_entries <- dims[1] * dims[2]
  for (i in 1:n_entries){
    if (A[i] != transpose[i]){
      bool <- FALSE
      break
    } else{
      bool <- TRUE
    }
}
return(bool)
}</pre>
```

## **Functionality Checks**

```
B <- matrix(c(-5, 2, 7, 1, 2, 3, 3, 6, -4), 3, 3)
C <- matrix(c(-5, 1, 3, 1, 2, 6, 3, 6, -4), 3, 3)
# results from my.is.symm()
my.is.symm(B)</pre>
```

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

my.is.symm(C)

## [1] TRUE