## 1 Review

```
model_mpg <- lm(mpg ~ hp + am + wt, data = mtcars)</pre>
summary(model_mpg)
##
## Call:
## lm(formula = mpg ~ hp + am + wt, data = mtcars)
##
## Residuals:
##
      Min
               1Q Median
                              ЗQ
                                     Max
## -3.4221 -1.7924 -0.3788 1.2249 5.5317
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 34.002875
                         2.642659 12.867 2.82e-13 ***
             2.083710
                                  1.514 0.141268
## am
                         1.376420
              -2.878575 0.904971 -3.181 0.003574 **
## wt
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.538 on 28 degrees of freedom
## Multiple R-squared: 0.8399, Adjusted R-squared: 0.8227
## F-statistic: 48.96 on 3 and 28 DF, p-value: 2.908e-11
anova(model_mpg) #anova stands for analysis of variance
## Analysis of Variance Table
##
## Response: mpg
            Df Sum Sq Mean Sq F value
                                        Pr(>F)
## hp
             1 678.37 678.37 105.354 5.395e-11 ***
## am
             1 202.24 202.24 31.408 5.335e-06 ***
             1 65.15
                      65.15 10.118 0.003574 **
## Residuals 28 180.29
                        6.44
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
```

```
678.37 + 202.24 + 65.15 + 180.29
## [1] 1126.05
mtcars %>%
 mutate(error = mpg - mean(mpg)) %>%
 mutate(error_sq = error^2) %>%
 glimpse()
## Observations: 32
## Variables: 13
             <dbl> 21.0, 21.0, 22.8, 21.4, 18.7, 18.1, 14.3, 24.4, 22.8,...
## $ mpg
## $ cyl
             <dbl> 6, 6, 4, 6, 8, 6, 8, 4, 4, 6, 6, 8, 8, 8, 8, 8, 8, 4,...
## $ disp
             <dbl> 160.0, 160.0, 108.0, 258.0, 360.0, 225.0, 360.0, 146....
## $ hp
             <dbl> 110, 110, 93, 110, 175, 105, 245, 62, 95, 123, 123, 1...
             <dbl> 3.90, 3.90, 3.85, 3.08, 3.15, 2.76, 3.21, 3.69, 3.92,...
## $ drat
## $ wt
             <dbl> 2.620, 2.875, 2.320, 3.215, 3.440, 3.460, 3.570, 3.19...
## $ qsec
             <dbl> 16.46, 17.02, 18.61, 19.44, 17.02, 20.22, 15.84, 20.0...
             <dbl> 0, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1,...
## $ vs
## $ am
             ## $ gear
             <dbl> 4, 4, 4, 3, 3, 3, 4, 4, 4, 4, 3, 3, 3, 3, 3, 3, 4,...
## $ carb
             <dbl> 4, 4, 1, 1, 2, 1, 4, 2, 2, 4, 4, 3, 3, 3, 4, 4, 4, 1,...
             <dbl> 0.909375, 0.909375, 2.709375, 1.309375, -1.390625, -1...
## $ error
## $ error_sq <dbl> 0.8269629, 0.8269629, 7.3407129, 1.7144629, 1.9338379...
mtcars %>%
 mutate(error = mpg - mean(mpg)) %>%
 mutate(error_sq = error^2) %>%
 summarize(sum(error_sq))
##
    sum(error_sq)
## 1
         1126.047
```

```
summary(model_mpg)$r.squared

## [1] 0.8398903
(678.37 + 202.24 + 65.15)/(678.37 + 202.24 + 65.15 + 180.29)

## [1] 0.8398917
```

## 2 Binomial Likelihood Example

You have a friend who claims that they can identify caffeinated and non-caffeniated coffee. Let their success rate be an unknown parameter p. You do a double blind test with 10 cups of coffee. Your friend identifies 8 cups of coffee correctly. Which of the following is the most likely estimation for p?

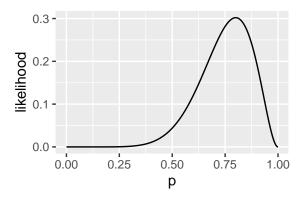
- a) 0.3
- b) 0.5
- c) 0.8

The answer may seem intuitive but let's see the reasoning behind it

```
p <- seq(from = 0, to = 1, by = 0.001)
likelihood <- dbinom(x = 8, size = 10, prob = p)

data<- as.data.frame(cbind(p, likelihood))

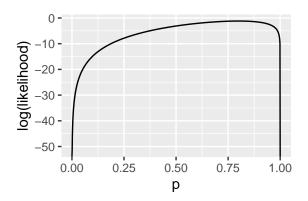
data %>%
    ggplot(aes(x = p, y = likelihood)) +
    geom_line()
```



```
data %>%
filter(likelihood == max(likelihood))

## p likelihood
## 1 0.8 0.3019899
```

```
data %>%
  ggplot(aes(x = p, y = log(likelihood))) +
  geom_line()
```



```
data %>%
filter(log(likelihood) == max(log(likelihood)))
## p likelihood
```

## 1 0.8 0.3019899

UCI - Stats 67 Fall 2019 Estimating p is quite easy when we are only testing one friend. We are now going to consider two scenarios. The experiment is repeated with a friend who tests 50 cups of coffee and correctly identifies 40 cups. What is the estimate for p? In second scenario: You repeat the experiment with five friends where each friend tests 10 cups of coffee. Two friends are correct about 7 cups, 1 friend is correct about 8 cups, 2 friends are correct about 9 cups. What is the estimate for p?

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