# **PSTAT 131 HW 2**

#### Ezra Torio

2022-10-16

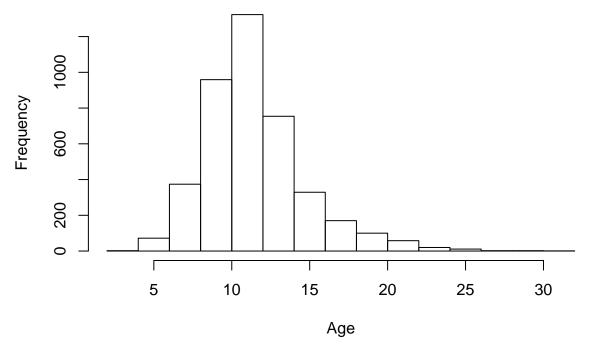
```
library(ggplot2)
library(tidyverse)
## -- Attaching packages -----
                                                ----- tidyverse 1.3.2 --
## v tibble 3.1.8 v dplyr 1.0.10
## v tidyr 1.2.0
                    v stringr 1.4.1
## v readr
           2.1.2
                     v forcats 0.5.2
          0.3.4
## v purrr
## -- Conflicts -----
                                  ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                   masks stats::lag()
library(tidymodels)
## -- Attaching packages ------ tidymodels 1.0.0 --
## v broom 1.0.1 v rsample ## v dials 1.0.0 v tune
                                     1.1.0
## v dials 1.0.0 v tune 1.0.0
## v infer 1.0.3 v workflows 1.1.0
## v modeldata 1.0.1 v workflowsets 1.0.0
               1.0.1
                      v yardstick 1.1.0
## v parsnip
## v recipes
               1.0.1
## -- Conflicts ------ tidymodels_conflicts() --
## x scales::discard() masks purrr::discard()
## x dplyr::filter() masks stats::filter()
## x recipes::fixed() masks stringr::fixed()
## x dplyr::lag() masks stats::lag()
## x yardstick::spec() masks readr::spec()
## x recipes::step() masks stats::step()
## * Learn how to get started at https://www.tidymodels.org/start/
library(corrplot)
## corrplot 0.92 loaded
library(ggthemes)
tidymodels_prefer()
getwd()
## [1] "/Users/ezratorio"
data <- read_csv("~/Desktop/abalone.csv")</pre>
## Rows: 4177 Columns: 9
## -- Column specification -----
## Delimiter: ","
```

```
## chr (1): type
## dbl (8): longest_shell, diameter, height, whole_weight, shucked_weight, visc...
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
head(data)
## # A tibble: 6 x 9
          longest_shell diameter height whole_weight shuck~1 visce~2 shell~3 rings
##
     <chr>
                    <dbl>
                             <dbl>
                                    <dbl>
                                                  <dbl>
                                                           <dbl>
                                                                   <dbl>
                                                                           <dbl> <dbl>
                             0.365
                                                         0.224
                                                                           0.15
## 1 M
                    0.455
                                    0.095
                                                  0.514
                                                                  0.101
                                                                                     15
## 2 M
                    0.35
                             0.265
                                    0.09
                                                  0.226
                                                         0.0995
                                                                 0.0485
                                                                           0.07
                                                                                      7
## 3 F
                    0.53
                             0.42
                                    0.135
                                                  0.677
                                                         0.256
                                                                  0.142
                                                                           0.21
                                                                                      9
## 4 M
                    0.44
                             0.365
                                    0.125
                                                         0.216
                                                                  0.114
                                                                           0.155
                                                                                     10
                                                  0.516
## 5 I
                    0.33
                             0.255
                                    0.08
                                                  0.205
                                                         0.0895
                                                                  0.0395
                                                                           0.055
                                                                                      7
## 6 I
                    0.425
                             0.3
                                    0.095
                                                  0.352 0.141
                                                                  0.0775
                                                                           0.12
                                                                                      8
    ... with abbreviated variable names 1: shucked_weight, 2: viscera_weight,
       3: shell_weight
```

#### Question 1

```
newData <- data
newData$age <- data$rings + 1.5
hist(newData$age, xlab = "Age", main = "Age Distribution of Abalones")</pre>
```

# **Age Distribution of Abalones**



The abalone ages are normally distributed and right skewed. Most abalones fall between 7 and 15 years old. It is very rare to find an abalone older than 20.

### Question 2

```
set.seed(823)
abalone_split <- initial_split(newData, prop = 0.80, strata = age)
abalone_train <- training(abalone_split)
abalone_test <- testing(abalone_split)</pre>
```

### Question 3

We should not use rings to predict age because age is just (rings +1.5) meaning that they would be perfectly correlated.

```
abalone_recipe <- recipe(age ~ ., data = abalone_train) %>%
  step_rm(rings) %>%
  step_dummy(all_nominal_predictors()) %>%
  step_interact(~ starts_with("type"):shucked_weight) %>%
  step_interact(~ longest_shell:diameter) %>%
   step_interact(~ shucked_weight:shell_weight) %>%
   step_normalize(all_predictors())
```

```
## Recipe
##
## Inputs:
##
##
         role #variables
##
      outcome
##
   predictor
##
## Operations:
##
## Variables removed rings
## Dummy variables from all_nominal_predictors()
## Interactions with starts_with("type"):shucked_weight
## Interactions with longest_shell:diameter
## Interactions with shucked weight:shell weight
## Centering and scaling for all_predictors()
```

#### Question 4

```
lm_model <- linear_reg() %>%
set_engine("lm")
```

## Question 5

```
lm_wflow <- workflow() %>%
  add_model(lm_model) %>%
  add_recipe(abalone_recipe)

lm_fit <- fit(lm_wflow, abalone_train)</pre>
```

### Question 6

```
testAbalone <- tibble(type = "F", longest_shell = 0.50, diameter = 0.10, height = 0.30,
                   whole_weight = 4, shucked_weight = 1, viscera_weight = 2,
                   shell_weight = 1, rings = 0)
predict(lm_fit, new_data = testAbalone)
## # A tibble: 1 x 1
##
     .pred
##
     <dbl>
## 1 23.2
Predicted age: 23.22974
Question 7
abalone_train_res <- predict(lm_fit, new_data = abalone_train %>% select(-age))
abalone_train_res %>%
 head()
## # A tibble: 6 x 1
##
     .pred
##
     <dbl>
## 1 9.47
## 2 8.11
## 3 9.77
## 4 10.4
## 5 10.1
abalone_train_res <- bind_cols(abalone_train_res, abalone_train %>% select(age))
abalone_train_res %>%
 head()
## # A tibble: 6 x 2
    .pred
##
            age
##
     <dbl> <dbl>
## 1 9.47
            8.5
## 2 8.11
           8.5
## 3 9.77
           8.5
## 4 10.4
             8.5
## 5 10.1
             9.5
## 6 6.28
             6.5
rmse(abalone_train_res, truth = age, estimate = .pred)
## # A tibble: 1 x 3
     .metric .estimator .estimate
##
     <chr>
           <chr>
                            <dbl>
## 1 rmse
             standard
                             2.17
abalone_metrics <- metric_set(rmse, rsq, mae)</pre>
abalone_metrics(abalone_train_res, truth = age,
                estimate = .pred)
## # A tibble: 3 x 3
```