

## Homework2

```
library(tidymodels)

## Warning: package 'tidymodels' was built under R version 4.0.5

## -- Attaching packages ----- tidymodels 0.2.0 --

## v broom      0.7.12   v recipes      0.2.0
## v dials      0.1.0    v rsample      0.1.1
## v dplyr      1.0.8    v tibble       3.1.6
## v ggplot2    3.3.5    v tidyverse    1.2.0
## v infer      1.0.0    v tune        0.2.0
## v modeldata   0.1.1    v workflows    0.2.6
## v parsnip     0.2.1    v workflowsets 0.2.1
## v purrr      0.3.4    v yardstick   0.0.9

## Warning: package 'broom' was built under R version 4.0.5

## Warning: package 'dials' was built under R version 4.0.5

## Warning: package 'dplyr' was built under R version 4.0.5

## Warning: package 'parsnip' was built under R version 4.0.5

## Warning: package 'recipes' was built under R version 4.0.5

## Warning: package 'tidyverse' was built under R version 4.0.5

## Warning: package 'tune' was built under R version 4.0.5

## Warning: package 'workflows' was built under R version 4.0.5

## Warning: package 'workflowsets' was built under R version 4.0.5

## -- Conflicts ----- tidymodels_conflicts() --
## x purrr::discard() masks scales::discard()
## x dplyr::filter()  masks stats::filter()
## x dplyr::lag()    masks stats::lag()
## x recipes::step() masks stats::step()
## * Use suppressPackageStartupMessages() to eliminate package startup messages
```

```
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.1 --

## v readr    2.1.2     vforcats 0.5.1
## v stringr 1.4.0

## Warning: package 'readr' was built under R version 4.0.5

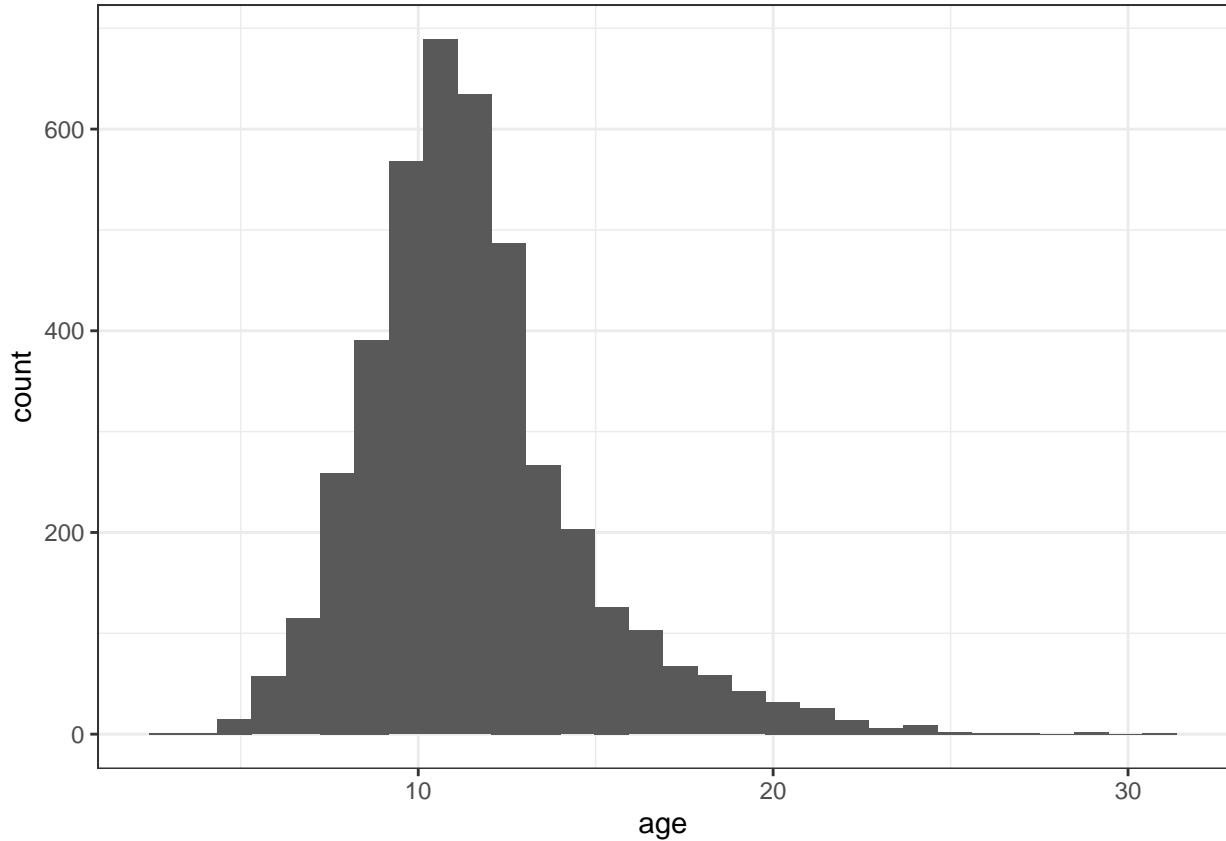
## -- Conflicts ----- tidyverse_conflicts() --
## x readr::col_factor() masks scales::col_factor()
## x purrrr::discard()   masks scales::discard()
## x dplyr::filter()     masks stats::filter()
## x stringr::fixed()    masks recipes::fixed()
## x dplyr::lag()        masks stats::lag()
## x readr::spec()       masks yardstick::spec()

abalone <- read.csv("abalone.csv")
```

## Question 1

```
# add age variable to the dataset
abalone <- abalone %>%
  mutate(age = rings + 1.5)

# make a histogram to assess the distribution of age
abalone %>%
  ggplot(aes(x=age)) +
  geom_histogram(bins=30) +
  theme_bw()
```



The distribution of age seems unsymmetrical and is a bit of a right skewed. Also, the most count age is around 11.

## Question 2

```
set.seed(1979)
abalone_split <- initial_split(abalone, prop = 0.8,
                                 strata = age)
abalone_train <- training(abalone_split)
abalone_test <- testing(abalone_split)
```

## Question 3

```
simple_abalone_recipe <-
  recipe(age ~ ., data = abalone_train %>% select(-rings)) %>%
  step_dummy(all_nominal_predictors()) %>%
  step_interact(~ shucked_weight:starts_with("type") + longest_shell:diameter + shucked_weight:shell_we
```

We should not use rings to predict age because we use rings to calculate age. When we use rings to predict age, our R^2 will become 1 even we do not include other predictors in our model. Therefore, our model is not meaningful when we include rings.

#### Question 4

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

#### Question 5

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

#### Question 6

```
lm_fit <- fit(lm_wflow, abalone_train)
```

```
lm_fit %>%  
  extract_fit_parsnip() %>%  
  tidy()
```

```
## # A tibble: 14 x 5  
##   term          estimate std.error statistic p.value  
##   <chr>        <dbl>     <dbl>     <dbl>     <dbl>  
## 1 (Intercept)    11.4      0.0370   309.       0  
## 2 longest_shell   0.743     0.282     2.63     8.49e- 3  
## 3 diameter        2.03      0.309     6.58     5.49e-11  
## 4 height          0.202     0.0683    2.96     3.12e- 3  
## 5 whole_weight     5.64      0.399     14.1     4.89e-44  
## 6 shucked_weight   -4.67     0.258    -18.1     6.37e-70  
## 7 viscera_weight   -1.19     0.158     -7.54    5.82e-14  
## 8 shell_weight      1.41      0.215     6.59     5.12e-11  
## 9 type_I          -0.897    0.116     -7.76    1.10e-14  
## 10 type_M         -0.177    0.103     -1.73    8.45e- 2  
## 11 shucked_weight_x_type_I    0.464    0.0867    5.36    8.96e- 8  
## 12 shucked_weight_x_type_M    0.214    0.107     1.99    4.61e- 2  
## 13 longest_shell_x_diameter   -2.84     0.404     -7.03    2.58e-12  
## 14 shucked_weight_x_shell_weight -0.0197  0.201     -0.0980  9.22e- 1  
  
abalone2 <- data.frame(type = 'F', longest_shell = 0.50, diameter = 0.10, height = 0.30, whole_weight = 0.15)  
predict_age <- predict(lm_fit, abalone2)  
predict_age  
  
## # A tibble: 1 x 1  
##   .pred  
##   <dbl>  
## 1 22.3
```

## Question 7

```
library(yardstick)
abalone_metrics <- metric_set(rmse,rsq,mae)

abalone_train_res <- predict(lm_fit, new_data = abalone_train %>% select(-rings,-age))
abalone_train_res %>% head()

## # A tibble: 6 x 1
##   .pred
##   <dbl>
## 1 9.56
## 2 8.09
## 3 9.82
## 4 10.1
## 5 10.9
## 6 6.24

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 9.56    8.5
## 2 8.09    8.5
## 3 9.82    8.5
## 4 10.1     9.5
## 5 10.9     9.5
## 6 6.24    6.5

abalone_metrics(abalone_train_res, truth = age,
                 estimate = .pred)

## # A tibble: 3 x 3
##   .metric .estimator .estimate
##   <chr>   <chr>        <dbl>
## 1 rmse    standard     2.14
## 2 rsq     standard    0.563
## 3 mae     standard    1.53
```

$R^2$  is 0.5633, which means 56.33% of the variability in  $Y(\text{age})$  that can be explained by using  $X$