

PSTAT 131 HW2

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```
library(tidymodels)

## -- Attaching packages ----- tidymodels 1.0.0 --
## v broom      1.0.1    v recipes      1.0.1
## v dials      1.0.0    v rsample     1.1.0
## v dplyr      1.0.9    v tibble     3.1.7
## v ggplot2    3.3.6    v tidyr      1.2.0
## v infer      1.0.3    v tune       1.0.0
## v modeldata  1.0.1    v workflows  1.1.0
## v parsnip    1.0.1    v workflowsets 1.0.0
## v purrr      0.3.4    v yardstick  1.1.0

## -- 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()
## * Search for functions across packages at https://www.tidymodels.org/find/

library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.2 --
## v readr      2.1.2    v forcats    0.5.2
## v stringr    1.4.0

## -- Conflicts ----- tidyverse_conflicts() --
## x readr::col_factor() masks scales::col_factor()
## x purrr::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()

f = "abalone.csv"

aba_data = read.csv(file = f)
head(aba_data)

##   type longest_shell diameter height whole_weight shucked_weight viscera_weight
## 1    M          0.455   0.365  0.095    0.5140         0.2245         0.1010
## 2    M          0.350   0.265  0.090    0.2255         0.0995         0.0485
## 3    F          0.530   0.420  0.135    0.6770         0.2565         0.1415
## 4    M          0.440   0.365  0.125    0.5160         0.2155         0.1140
## 5    I          0.330   0.255  0.080    0.2050         0.0895         0.0395
## 6    I          0.425   0.300  0.095    0.3515         0.1410         0.0775
```

```
##   shell_weight rings
## 1      0.150    15
## 2      0.070     7
## 3      0.210     9
## 4      0.155    10
## 5      0.055     7
## 6      0.120     8
```

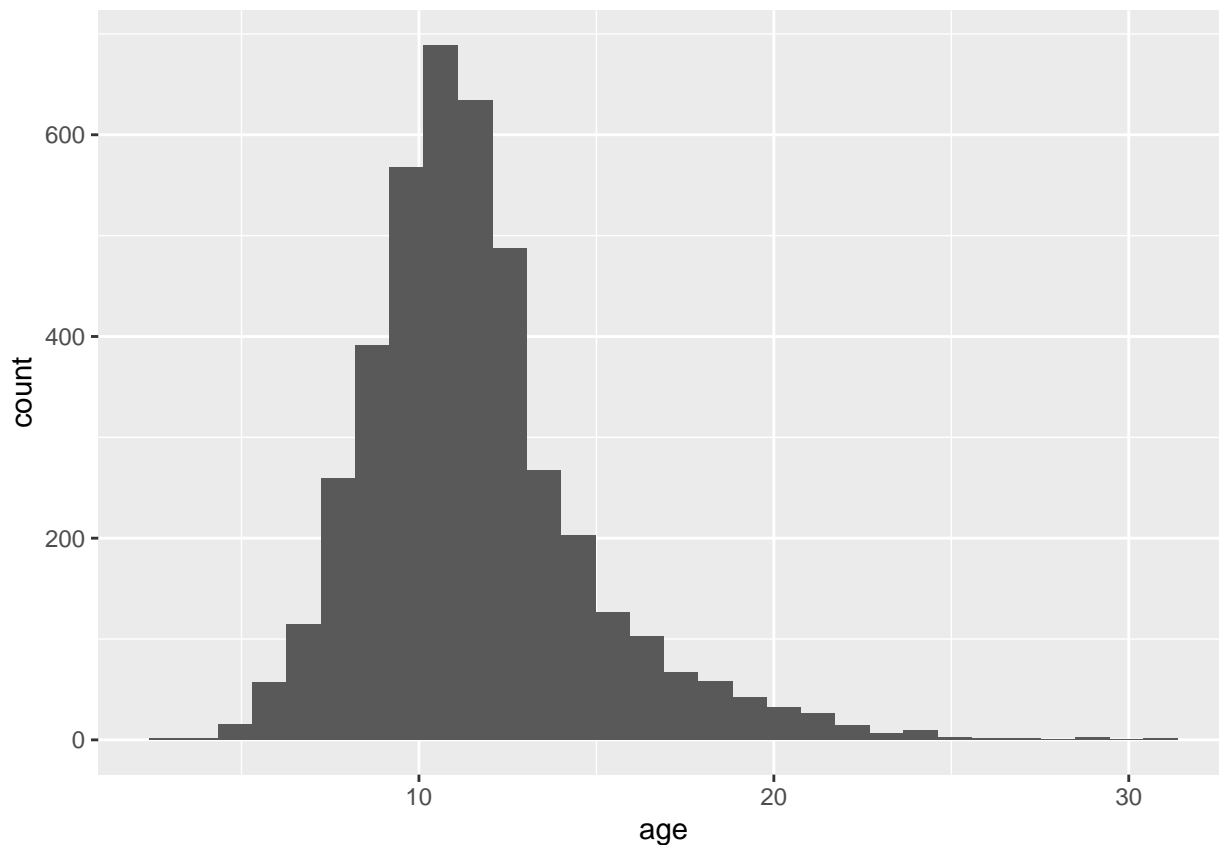
Question 1

```
aba_data$age <- aba_data$rings + 1.5
head(aba_data)
```

```
##   type longest_shell diameter height whole_weight shucked_weight viscera_weight
## 1    M      0.455      0.365  0.095      0.5140      0.2245      0.1010
## 2    M      0.350      0.265  0.090      0.2255      0.0995      0.0485
## 3    F      0.530      0.420  0.135      0.6770      0.2565      0.1415
## 4    M      0.440      0.365  0.125      0.5160      0.2155      0.1140
## 5    I      0.330      0.255  0.080      0.2050      0.0895      0.0395
## 6    I      0.425      0.300  0.095      0.3515      0.1410      0.0775
##   shell_weight rings  age
## 1      0.150     15 16.5
## 2      0.070      7  8.5
## 3      0.210      9 10.5
## 4      0.155     10 11.5
## 5      0.055      7  8.5
## 6      0.120      8  9.5
```

```
aba_data%>%
  ggplot(aes(x = age)) +
  geom_histogram()
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



The distribution of age is left skewed, much of the mass of its distribution is at the lower end, majority of the abalones are aged less than 15.

Question 2

```
set.seed(1000)

aba_split <- initial_split(aba_data, prop = 0.75,
                           strata = age)
aba_train <- training(aba_split)
aba_test  <- testing(aba_split)
```

Question 3

Rings should not be included because the age is directly calculated from rings. There is a super strong correlation between the two variables. If rings is included in the predictors, the model will be overfit.

```
aba_recipe <-
  recipe(age ~ type + longest_shell + diameter + height + whole_weight + shucked_weight + viscera_weight) %>%
  step_dummy(all_nominal_predictors()) %>%
  step_interact(terms = ~ starts_with('type'):shucked_weight) %>%
  step_interact(terms = ~ longest_shell:diameter) %>%
  step_interact(terms = ~ shucked_weight:shell_weight) %>%
  step_center(all_predictors()) %>%
  step_scale(all_predictors())
aba_recipe
```

```
## Recipe
##
```

```
## Inputs:
##
##      role #variables
##      outcome      1
##      predictor      8
##
## Operations:
##
## 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 for all_predictors()
## Scaling for all_predictors()
```

Question 4

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

Question 5

```
aba_lm_wflow <- workflow() %>%
  add_model(lm_model) %>%
  add_recipe(aba_recipe)
```

Question 6

```
lm_fit <- fit(aba_lm_wflow, aba_train)
lm_fit
```

```
## == Workflow [trained] =====
## Preprocessor: Recipe
## Model: linear_reg()
##
## -- Preprocessor -----
## 6 Recipe Steps
##
## * step_dummy()
## * step_interact()
## * step_interact()
## * step_interact()
## * step_center()
## * step_scale()
##
## -- Model -----
##
## Call:
## stats::lm(formula = ..y ~ ., data = data)
##
## Coefficients:
##              (Intercept)              longest_shell
##              11.42558              0.27720
##              diameter              height
##              2.26024              0.24242
##              whole_weight      shucked_weight
##              5.19864              -4.51219
```

```
##          viscera_weight          shell_weight
##          -1.06043          1.43975
##          type_I          type_M
##          -0.95225          -0.32397
##          type_I_x_shucked_weight          type_M_x_shucked_weight
##          0.51578          0.38995
##          longest_shell_x_diameter          shucked_weight_x_shell_weight
##          -2.56822          -0.04577
```

```
predict(lm_fit, data.frame(type = "F", longest_shell = 0.50, diameter = 0.10, height = 0.30, whole_weight
```

```
## # A tibble: 1 x 1
##   .pred
##   <dbl>
## 1  21.1
```

Question 7

```
library(yardstick)
aba_metric = metric_set(rsq, rmse, mae)
aba_train_result = predict(lm_fit, aba_train %>% select(-age, -rings))
aba_train_result = bind_cols(aba_train_result, aba_train %>% select(age))
head(aba_train_result)
```

```
## # A tibble: 6 x 2
##   .pred age
##   <dbl> <dbl>
## 1  8.15  8.5
## 2  9.34  9.5
## 3 10.4   8.5
## 4 10.0   9.5
## 5 11.0   9.5
## 6  6.35  6.5
```

```
aba_metrics = aba_metric(aba_train_result, truth = age, estimate = .pred)
aba_metrics
```

```
## # A tibble: 3 x 3
##   .metric .estimator .estimate
##   <chr>    <chr>         <dbl>
## 1 rsq      standard      0.558
## 2 rmse     standard      2.15
## 3 mae      standard      1.55
```