# ADA2: Class 11, Chs 05 and 07, writing and plotting model equations

Code **▼** 

#### **Name Here**

#### February 25, 2022

- 11. Two-way main-effect model: Kangaroo crest width
  - 1.1 (3 p) Write the fitted model equation.
    - 1.1.1 Solution
  - 1.2 (2 p) Separate model equations.
    - 1.2.1 Solution
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- 22. ANCOVA model: Faculty political tolerances
  - 2.1 (3 p) Write the fitted model equation.
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#### This assignment is to be printed and hand-written.

In my opinion, some of the most important skills in modeling are:

- writing down a model using indicator variables,
- interpretting model coefficients,
- solving for the predicted value for any combination of predictors, and
- plotting the fitted model.

This assignment applies these skills to two-way factor models (ADA2 Chapter 5) and ANCOVA models with one factor and one continuous predictor (ADA2 Chapter 7).

# 11. Two-way main-effect model: Kangaroo crest width

Recall these data, results, and the model from Week 05.

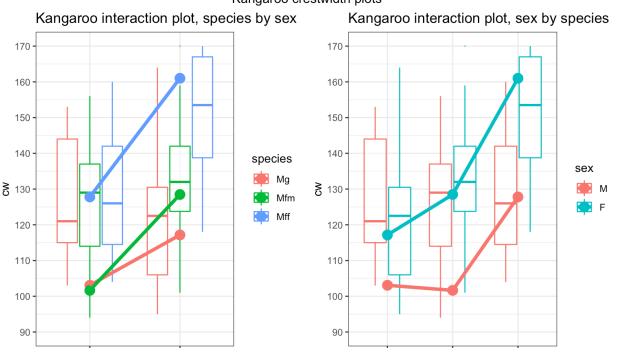
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```
library(erikmisc)
library(tidyverse)
kang <-
 read_csv(
    "~/Dropbox/3_Education/Courses/stat_528_ada2/ADA2_CL_09_kang.csv"
 , na = c("", ".")
 ) %>%
 # subset only our columns of interest
    sex, species, cw
 ) %>%
 # make dose a factor variable and label the levels
 mutate(
                            , labels = c("M","F"))
   sex
            = factor(sex
   species = factor(species, labels = c("Mg", "Mfm", "Mff"))
str(kang)
```

```
tibble [148 × 3] (S3: tbl_df/tbl/data.frame)
$ sex : Factor w/ 2 levels "M","F": 1 1 1 1 1 1 1 1 1 1 1 1 1 ...
$ species: Factor w/ 3 levels "Mg","Mfm","Mff": 1 1 1 1 1 1 1 1 1 1 1 ...
$ cw : num [1:148] 153 141 144 116 120 188 149 128 151 103 ...
```

```
# A tibble: 6 \times 3
         species
  sex
  <fct> <fct>
                  <dbl>
1 M
         Mg
                   103.
2 M
         Mfm
                   102.
3 M
         Mff
                   128.
4 F
        Mg
                   117.
5 F
                   128.
         Mfm
6 F
         Mff
                   161
```

#### Kangaroo crestwidth plots



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M F Mg Mfm Mff sex species

```
lm_cw_x_s <-
lm(
    cw ~ sex + species
, data = kang
)
# parameter estimate table
summary(lm_cw_x_s)</pre>
```

```
Call:
lm(formula = cw \sim sex + species, data = kang)
Residuals:
   Min
            10 Median
-94.456 -19.746 1.553 23.478 90.216
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
                         6.039 16.193 < 2e-16 ***
(Intercept)
            97.784
                         6.070 4.064 7.89e-05 ***
             24.673
speciesMfm
              4.991
                         7.460
                                 0.669
                                          0.505
             34.280
                         7.383
                                 4.643 7.66e-06 ***
speciesMff
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
Residual standard error: 36.91 on 144 degrees of freedom
Multiple R-squared: 0.2229,
                              Adjusted R-squared: 0.2067
F-statistic: 13.77 on 3 and 144 DF, p-value: 6.057e-08
```

## 1.1 (3 p) Write the fitted model equation.

Use the parameter estimate table above to write out the fitted model equation. Use indicator function notation for categorical variables. First determine what each sex and species number is. The equation looks like:  $\hat{y} = [\text{terms}]$ .

#### 1.1.1 Solution

$$\widehat{cw} = 97.8 + 24.7 * I(\text{sex} = \text{F}) + 5.0 * I(\text{sp.} = \text{Mfm}) + 34.3 * I(\text{sp.} = \text{Mff})$$

## 1.2 (2 p) Separate model equations.

For each combination of species and sex, write the model.

#### 1.2.1 Solution

Sex	Species	Fitted Model
М	Mg	$\hat{y} = 97.8$
\		
М	Mfm	$\hat{y} = 97.8 + 5.0 = 102.8$

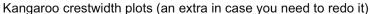
Sex	Species	Fitted Model
\		
М	Mff	$\hat{y} = 97.8 + 34.3 = 132.1$
\		
F	Mg	\$= 97.8 + 24.7 = 122.5 \$
\		
F	Mfm	$\hat{y} = 97.8 + 24.7 + 5.0 = 127.5$
\		
F	Mff	$\hat{y} = 97.8 + 24.7 + 34.3 = 156.8$

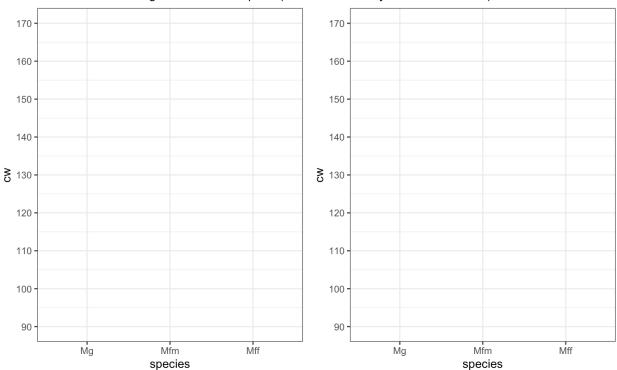
## 1.3 (0 p) Plot the observed and fitted values.

Use symbols/colors/labels to distinguish between the observed and predicted values and clearly identify the species/sex combinations. Use the minimum about of labeling to make it clear.

## 1.3.1 Solution

COVID-19 Year, no hand plotting:(





# 2 2. ANCOVA model: Faculty political tolerances

A political scientist developed a questionnaire to determine political tolerance scores for a random sample of faculty members at her university. She wanted to compare mean scores adjusted for the age for each of the three categories: full professors (coded 1), associate professors (coded 2), and assistant professors (coded 3). The data are given below. Note the higher the score, the more tolerant the individual.

Below we will fit and interpret a model to assess the dependence of tolerance score on age and rank. (We will assess model fit in a later

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assignment.)

```
tolerate <-
    read_csv("~/Dropbox/3_Education/Courses/stat_528_ada2/ADA2_CL_12_tolerate.csv") %>%
    mutate(
        rank = factor(rank)
        # set "3" as baseline level
        , rank = relevel(rank, "3")
        )
    str(tolerate)
```

```
spec_tbl_df [30 x 3] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
$ score: num [1:30] 3.03 4.31 5.09 3.71 5.29 2.7 2.7 4.02 5.52 4.62 ...
$ age : num [1:30] 65 47 49 41 40 61 52 45 41 39 ...
$ rank : Factor w/ 3 levels "3","1","2": 2 2 2 2 2 2 2 2 2 2 2 ...
- attr(*, "spec")=
    .. cols(
    .. score = col_double(),
    .. age = col_double(),
    .. rank = col_double()
    .. )
- attr(*, "problems")=<externalptr>
```

## 2.1 (3 p) Write the fitted model equation.

Note in the code what the baseline rank is.

```
lm_s_a_r_ar <-
lm(
    score ~ age * rank
    , data = tolerate
    )
summary(lm_s_a_r_ar)</pre>
```

```
Call:
lm(formula = score ~ age * rank, data = tolerate)
Residuals:
              1Q
                   Median
    Min
                                30
                                        Max
-1.34746 -0.28793 0.01405 0.36653 1.07669
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 5.42706
                       0.98483
                                 5.511 1.15e-05 ***
           -0.01321
age
                       0.02948 -0.448
                                         0.6580
rank1
            2.78490
                       1.51591
                                 1.837
                                         0.0786 .
rank2
           -1.22343
                       1.50993 -0.810
                                         0.4258
           -0.07247
                       0.03779 -1.918
age:rank1
                                         0.0671 .
age:rank2
            0.03022
                       0.04165
                                 0.726
                                         0.4751
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.6378 on 24 degrees of freedom
Multiple R-squared: 0.5112,
                               Adjusted R-squared: 0.4093
F-statistic: 5.02 on 5 and 24 DF, p-value: 0.002748
```

Use the parameter estimate table above to write out the fitted model equation. Use indicator function notation for categorical variables. The equation looks like:  $\hat{y} = [\text{terms}]$ .

### 2.1.1 Solution

```
\widehat{score} = 5.4 - 0.013 * Age + 2.785 * I(rank = 1) - 1.223 * I(rank = 2) - 0.072 * Age * I(rank = 1) + 0.030 * Age * I(rank = 2)
```

## 2.2 (2 p) Separate model equations.

There's a separate regression line for each faculty rank.

### 2.2.1 Solution

Rank	Fitted Model		
1	$\hat{y} = 5.4 - 0.013 * Age + 2.785 - 0.072 * Age = 8.1850.085 * Age$		
\			
2	\$= 5.4 - 0.013 * Age - 1.223 + 0.030 * Age = 4.177 + 0.017 * Age \$		
\			
3	$\hat{y} = 5.4 - 0.013 * Age$		

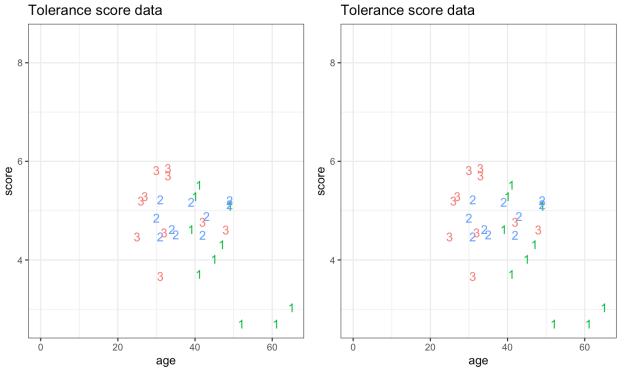
## 2.3 (0 p) Plot the fitted regression lines.

Use symbols/colors/labels to distinguish between the observed and predicted values and clearly identify the rank. Use the minimum about of labeling to make it clear. I recommend plotting each line by evaluating two points then connecting them, for example, by evaluating at age=0 and age=50.

#### 2.3.1 Solution

COVID-19 Year, no hand plotting:(

## Faculty tolerance (an extra in case you need to redo it)



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