

Coding Challenge 7

Theresa Quintana

2025-04-03

You can find the corresponding documents to this assignment on my GitHub: Theresa's PLPA 6820 Github Access

#Question 1. 4 pts. Read in the data called "PlantEmergence.csv" using a relative file path and load the following libraries. tidyverse, lme4, emmeans, multcomp, and multcompView. Turn the Treatment , DaysAfterPlanting and Rep into factors using the function as.factor

```
library(tidyverse)
library(lme4)
library(emmeans)
library(multcomp)
library(multcompView)

emergence <- read.csv("PlantEmergence.csv", na.strings = "na")

emergence$Treatment <- as.factor(emergence$Treatment)
emergence$DaysAfterPlanting <- as.factor(emergence$DaysAfterPlanting)
emergence$Rep <- as.factor(emergence$Rep)
```

#Question 2. 5 pts. Fit a linear model to predict Emergence using Treatment and DaysAfterPlanting along with the interaction. Provide the summary of the linear model and ANOVA results.

```
#Effect of one variable depends on the other; use the * sign to include the interaction
lm.interaction <- lm(Emergence ~ Treatment*DaysAfterPlanting, data = emergence)
summary(lm.interaction)
```

```
##
## Call:
## lm(formula = Emergence ~ Treatment * DaysAfterPlanting, data = emergence)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -21.250  -6.062  -0.875   6.750  21.875
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   1.823e+02  5.324e+00  34.229  <2e-16 ***
## Treatment2   -1.365e+02  7.530e+00 -18.128  <2e-16 ***
## Treatment3    1.112e+01  7.530e+00   1.477    0.142
## Treatment4    2.500e+00  7.530e+00   0.332    0.741
```

```

## Treatment5          8.750e+00  7.530e+00  1.162  0.248
## Treatment6          7.000e+00  7.530e+00  0.930  0.355
## Treatment7         -1.250e-01  7.530e+00 -0.017  0.987
## Treatment8          9.125e+00  7.530e+00  1.212  0.228
## Treatment9          2.375e+00  7.530e+00  0.315  0.753
## DaysAfterPlanting14  1.000e+01  7.530e+00  1.328  0.187
## DaysAfterPlanting21  1.062e+01  7.530e+00  1.411  0.161
## DaysAfterPlanting28  1.100e+01  7.530e+00  1.461  0.147
## Treatment2:DaysAfterPlanting14  1.625e+00  1.065e+01  0.153  0.879
## Treatment3:DaysAfterPlanting14 -2.625e+00  1.065e+01 -0.247  0.806
## Treatment4:DaysAfterPlanting14 -6.250e-01  1.065e+01 -0.059  0.953
## Treatment5:DaysAfterPlanting14  2.500e+00  1.065e+01  0.235  0.815
## Treatment6:DaysAfterPlanting14  1.000e+00  1.065e+01  0.094  0.925
## Treatment7:DaysAfterPlanting14 -2.500e+00  1.065e+01 -0.235  0.815
## Treatment8:DaysAfterPlanting14 -2.500e+00  1.065e+01 -0.235  0.815
## Treatment9:DaysAfterPlanting14  6.250e-01  1.065e+01  0.059  0.953
## Treatment2:DaysAfterPlanting21  3.500e+00  1.065e+01  0.329  0.743
## Treatment3:DaysAfterPlanting21 -1.000e+00  1.065e+01 -0.094  0.925
## Treatment4:DaysAfterPlanting21  1.500e+00  1.065e+01  0.141  0.888
## Treatment5:DaysAfterPlanting21  2.875e+00  1.065e+01  0.270  0.788
## Treatment6:DaysAfterPlanting21  4.125e+00  1.065e+01  0.387  0.699
## Treatment7:DaysAfterPlanting21 -2.125e+00  1.065e+01 -0.200  0.842
## Treatment8:DaysAfterPlanting21 -1.500e+00  1.065e+01 -0.141  0.888
## Treatment9:DaysAfterPlanting21 -1.250e+00  1.065e+01 -0.117  0.907
## Treatment2:DaysAfterPlanting28  2.750e+00  1.065e+01  0.258  0.797
## Treatment3:DaysAfterPlanting28 -1.875e+00  1.065e+01 -0.176  0.861
## Treatment4:DaysAfterPlanting28  3.264e-13  1.065e+01  0.000  1.000
## Treatment5:DaysAfterPlanting28  2.500e+00  1.065e+01  0.235  0.815
## Treatment6:DaysAfterPlanting28  2.125e+00  1.065e+01  0.200  0.842
## Treatment7:DaysAfterPlanting28 -3.625e+00  1.065e+01 -0.340  0.734
## Treatment8:DaysAfterPlanting28 -1.500e+00  1.065e+01 -0.141  0.888
## Treatment9:DaysAfterPlanting28 -8.750e-01  1.065e+01 -0.082  0.935
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10.65 on 108 degrees of freedom
## Multiple R-squared:  0.9585, Adjusted R-squared:  0.945
## F-statistic: 71.21 on 35 and 108 DF,  p-value: < 2.2e-16

```

```
anova(lm.interaction)
```

```

## Analysis of Variance Table
##
## Response: Emergence
##
##           Df Sum Sq Mean Sq F value    Pr(>F)
## Treatment      8 279366   34921 307.9516 < 2.2e-16 ***
## DaysAfterPlanting  3   3116    1039  9.1603 1.877e-05 ***
## Treatment:DaysAfterPlanting 24    142     6  0.0522      1
## Residuals     108  12247    113
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

#Question 3. 5 pts. Based on the results of the linear model in question 2, do you need to fit the interaction

term? No, because the p-value for the interaction (Treatment:DaysAfterPlanting) is 1, which is > 0.05 and so is not significant.

#Provide a simplified linear model without the interaction term but still testing both main effects. Provide the summary and ANOVA results. Then, interpret the intercept and the coefficient for Treatment 2.

#Interpretation from below code (simplified linear model) = Intercept is 182.163, so emergence is 182.163 with treatment one at 7 days after planting. The coefficient for Treatment 2 is -134.531, which means a decrease in emergence (y/dependent variable) of 134.531 when treatment 2 is applied compared to treatment 1, holding days after planting constant*

#simplified = each variable impacts outcome, but they do not influence each others effect (independent&
simplified.lm <- lm(Emergence ~ Treatment + DaysAfterPlanting, data = emergence)

```
summary(simplified.lm)
```

```
##
## Call:
## lm(formula = Emergence ~ Treatment + DaysAfterPlanting, data = emergence)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -21.1632  -6.1536  -0.8542   6.1823  21.3958
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      182.163      2.797   65.136 < 2e-16 ***
## Treatment2      -134.531      3.425  -39.277 < 2e-16 ***
## Treatment3         9.750      3.425   2.847  0.00513 **
## Treatment4         2.719      3.425   0.794  0.42876
## Treatment5        10.719      3.425   3.129  0.00216 **
## Treatment6         8.812      3.425   2.573  0.01119 *
## Treatment7        -2.188      3.425  -0.639  0.52416
## Treatment8         7.750      3.425   2.263  0.02529 *
## Treatment9         2.000      3.425   0.584  0.56028
## DaysAfterPlanting14  9.722      2.283   4.258 3.89e-05 ***
## DaysAfterPlanting21 11.306      2.283   4.951 2.21e-06 ***
## DaysAfterPlanting28 10.944      2.283   4.793 4.36e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 9.688 on 132 degrees of freedom
## Multiple R-squared:  0.958, Adjusted R-squared:  0.9545
## F-statistic: 273.6 on 11 and 132 DF, p-value: < 2.2e-16
```

```
anova(simplified.lm)
```

```
## Analysis of Variance Table
##
## Response: Emergence
##              Df Sum Sq Mean Sq F value    Pr(>F)
## Treatment      8 279366   34921 372.070 < 2.2e-16 ***
## DaysAfterPlanting 3   3116    1039 11.068 1.575e-06 ***
## Residuals     132 12389     94
```

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

#Question 4. 5 pts. Calculate the least square means for Treatment using the emmeans package and perform a Tukey separation with the compact letter display using the cld function.

#Interpret the results - First we are calculating the average emergence for each treatment (controlling days after planting). Treatment 2 has the lowest emergence, while the rest look rather uniform. Next, we use the cld to group the treatments to compared least square means for statistical differences. Treatments in the same or overlapping groups are not significantly different from one another ($p > 0.05$). The group numbers start at the lowest emergence (1) and increase (here the highest is 3). Treatment 2 is in its own group (significantly different from the rest) and the lowest emergence(1).

```
lsmeans <- emmeans(simplified.lm, ~Treatment)
lsmeans
```

```
## Treatment emmean SE df lower.CL upper.CL
## 1          190.2 2.42 132    185.4    194.9
## 2           55.6 2.42 132     50.8     60.4
## 3          199.9 2.42 132    195.1    204.7
## 4          192.9 2.42 132    188.1    197.7
## 5          200.9 2.42 132    196.1    205.7
## 6          199.0 2.42 132    194.2    203.8
## 7          188.0 2.42 132    183.2    192.8
## 8          197.9 2.42 132    193.1    202.7
## 9          192.2 2.42 132    187.4    196.9
##
## Results are averaged over the levels of: DaysAfterPlanting
## Confidence level used: 0.95
```

```
results_lsmeans <- cld(lsmeans, alpha = 0.05, details = TRUE)
results_lsmeans
```

```
## $emmeans
## Treatment emmean SE df lower.CL upper.CL .group
## 2          55.6 2.42 132     50.8     60.4    1
## 7          188.0 2.42 132    183.2    192.8    2
## 1          190.2 2.42 132    185.4    194.9    23
## 9          192.2 2.42 132    187.4    196.9    23
## 4          192.9 2.42 132    188.1    197.7    23
## 8          197.9 2.42 132    193.1    202.7    23
## 6          199.0 2.42 132    194.2    203.8    3
## 3          199.9 2.42 132    195.1    204.7    3
## 5          200.9 2.42 132    196.1    205.7    3
##
## Results are averaged over the levels of: DaysAfterPlanting
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 9 estimates
## significance level used: alpha = 0.05
## NOTE: If two or more means share the same grouping symbol,
##       then we cannot show them to be different.
##       But we also did not show them to be the same.
##
```

```
## $comparisons
## contrast estimate SE df t.ratio p.value
## Treatment7 - Treatment2 132.344 3.43 132 38.638 <.0001
## Treatment1 - Treatment2 134.531 3.43 132 39.277 <.0001
## Treatment1 - Treatment7 2.188 3.43 132 0.639 0.9993
## Treatment9 - Treatment2 136.531 3.43 132 39.861 <.0001
## Treatment9 - Treatment7 4.188 3.43 132 1.223 0.9502
## Treatment9 - Treatment1 2.000 3.43 132 0.584 0.9997
## Treatment4 - Treatment2 137.250 3.43 132 40.071 <.0001
## Treatment4 - Treatment7 4.906 3.43 132 1.432 0.8832
## Treatment4 - Treatment1 2.719 3.43 132 0.794 0.9969
## Treatment4 - Treatment9 0.719 3.43 132 0.210 1.0000
## Treatment8 - Treatment2 142.281 3.43 132 41.540 <.0001
## Treatment8 - Treatment7 9.938 3.43 132 2.901 0.0978
## Treatment8 - Treatment1 7.750 3.43 132 2.263 0.3724
## Treatment8 - Treatment9 5.750 3.43 132 1.679 0.7583
## Treatment8 - Treatment4 5.031 3.43 132 1.469 0.8678
## Treatment6 - Treatment2 143.344 3.43 132 41.850 <.0001
## Treatment6 - Treatment7 11.000 3.43 132 3.212 0.0425
## Treatment6 - Treatment1 8.812 3.43 132 2.573 0.2083
## Treatment6 - Treatment9 6.812 3.43 132 1.989 0.5538
## Treatment6 - Treatment4 6.094 3.43 132 1.779 0.6957
## Treatment6 - Treatment8 1.062 3.43 132 0.310 1.0000
## Treatment3 - Treatment2 144.281 3.43 132 42.124 <.0001
## Treatment3 - Treatment7 11.938 3.43 132 3.485 0.0187
## Treatment3 - Treatment1 9.750 3.43 132 2.847 0.1120
## Treatment3 - Treatment9 7.750 3.43 132 2.263 0.3724
## Treatment3 - Treatment4 7.031 3.43 132 2.053 0.5099
## Treatment3 - Treatment8 2.000 3.43 132 0.584 0.9997
## Treatment3 - Treatment6 0.938 3.43 132 0.274 1.0000
## Treatment5 - Treatment2 145.250 3.43 132 42.406 <.0001
## Treatment5 - Treatment7 12.906 3.43 132 3.768 0.0074
## Treatment5 - Treatment1 10.719 3.43 132 3.129 0.0535
## Treatment5 - Treatment9 8.719 3.43 132 2.545 0.2204
## Treatment5 - Treatment4 8.000 3.43 132 2.336 0.3288
## Treatment5 - Treatment8 2.969 3.43 132 0.867 0.9943
## Treatment5 - Treatment6 1.906 3.43 132 0.557 0.9998
## Treatment5 - Treatment3 0.969 3.43 132 0.283 1.0000
##
## Results are averaged over the levels of: DaysAfterPlanting
## P value adjustment: tukey method for comparing a family of 9 estimates
```

#Question 5. 4 pts. The provided function lets you dynamically add a linear model plus one factor from that model and plots a bar chart with letters denoting treatment differences. Use this model to generate the plot shown below.

#Explain the significance of the letters - The letters have the same meaning as the numbers in the previous question. Treatments in the same or overlapping groups are not significantly different from one another ($p > 0.05$). The group numbers start at the highest emergence (a) and decrease (here the lowest is c). Treatment 2 is in its own group (significantly different from the rest) and the lowest emergence (c).

```
plot_cldbars_onefactor <- function(lm_model, factor) {
  data <- lm_model$model
  variables <- colnames(lm_model$model)
```

```

dependent_var <- variables[1]
independent_var <- variables[2:length(variables)]

lsmeans <- emmeans(lm_model, as.formula(paste("~", factor))) # estimate lsmeans
Results_lsmeans <- cld(lsmeans, alpha = 0.05, reversed = TRUE, details = TRUE, Letters = letters) # c

# Extracting the letters for the bars
sig.diff.letters <- data.frame(Results_lsmeans$emmeans[,1],
                              str_trim(Results_lsmeans$emmeans[,7]))
colnames(sig.diff.letters) <- c(factor, "Letters")

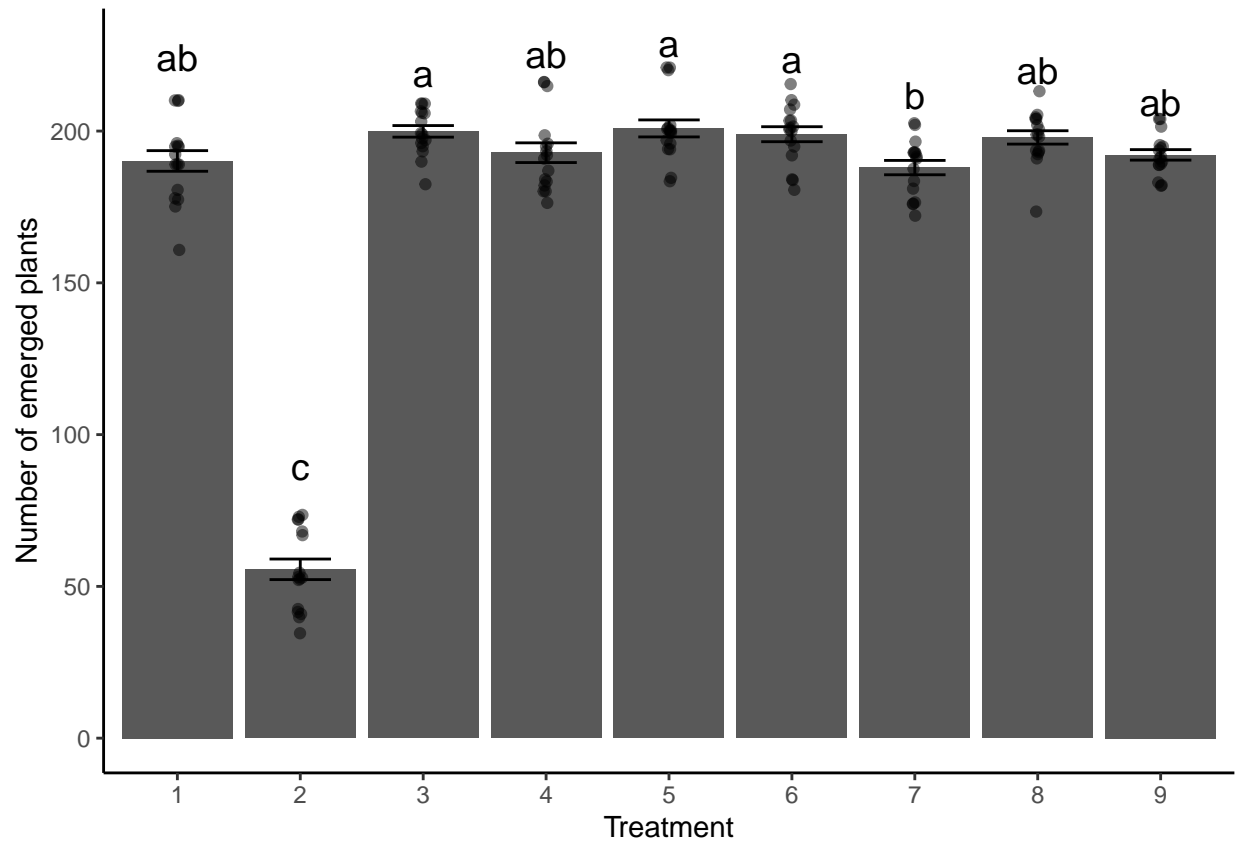
# for plotting with letters from significance test
ave_stand2 <- lm_model$model %>%
  group_by(!sym(factor)) %>%
  dplyr::summarize(
    ave.emerge = mean(.data[[dependent_var]], na.rm = TRUE),
    se = sd(.data[[dependent_var]]) / sqrt(n())
  ) %>%
  left_join(sig.diff.letters, by = factor) %>%
  mutate(letter_position = ave.emerge + 10 * se)

plot <- ggplot(data, aes(x = !! sym(factor), y = !! sym(dependent_var))) +
  stat_summary(fun = mean, geom = "bar") +
  stat_summary(fun.data = mean_se, geom = "errorbar", width = 0.5) +
  ylab("Number of emerged plants") +
  geom_jitter(width = 0.02, alpha = 0.5) +
  geom_text(data = ave_stand2, aes(label = Letters, y = letter_position), size = 5) +
  xlab(as.character(factor)) +
  theme_classic()

return(plot)
}

plot_cldbars_onefactor(simplified.lm, "Treatment")

```



#Question 6. 2 pts. Generate the gfm .md file along with a .html, .docx, or .pdf. Commit, and push the .md file to github and turn in the .html, .docx, or .pdf to Canvas. Provide me a link here to your github.

#Link to github is a clickable link at the top of the document.