CviMVP_Nursery_Hatchery_Analysis

2024-02-22

R Markdown

```
#load necessary packages
library(tidyverse)
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr 1.1.4
                                  2.1.5
                       v readr
## v forcats 1.0.0
                       v stringr
                                   1.5.1
## v ggplot2 3.5.1
                    v tibble
                                   3.2.1
## v lubridate 1.9.3
                    v tidyr
                                   1.3.1
## v purrr
              1.0.2
## -- Conflicts -----
                                         ## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                   masks stats::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
library(ggplot2)
library(gridExtra)
##
## Attaching package: 'gridExtra'
## The following object is masked from 'package:dplyr':
##
##
      combine
library(multcompView)
library(car) #upload package with Anova function
## Loading required package: carData
##
## Attaching package: 'car'
## The following object is masked from 'package:dplyr':
##
##
      recode
##
## The following object is masked from 'package:purrr':
##
##
      some
```

```
library(cowplot)
##
## Attaching package: 'cowplot'
## The following object is masked from 'package:lubridate':
##
##
       stamp
#Set working directory
setwd(dirname(rstudioapi::getActiveDocumentContext()$path))
#download data in csv format
#download spawn treatment data
spawn <- read.csv("../../data/performance_H2F/CviMVP_Larvae_Growth-spawn_trt.csv")</pre>
#download group data
group <- read.csv("../../data/performance_H2F/CviMVP_Larvae_Growth-group.csv")</pre>
#download larvae image length data
larvae_img <- read.csv(".../..data/performance_H2F/CviMVP_Larvae_Growth-larvae_img.csv")</pre>
#make mega sheet with all length data
#join image data to group data in new dataframe
len_df <-group %>%
          left_join(larvae_img, by = "Group_Day_Key")
#remove some columns
len_analysis_df <- subset(len_df, select = -c(Ind_ID, Group_Day_Label.y, Image_Key, Img_Notes,SpawnTrt_)</pre>
#rename columns
#subset data for oysters aged 15-78 days (later stage offspring)
latestage_df <- subset(len_analysis_df, Day_analysis == "78" | Day_analysis == "15-21")
#a sample marked WV is erroneously in this data set- remove
latestage_df <- subset(latestage_df, Image_Label != "WV_06_200L_6")</pre>
#rename cols
colnames(latestage_df) <- c("site_name", "site_day_key", "site_day_label", "day_trt", "day_an", "day",</pre>
#make column with measurements in millimeters instead of microns
latestage_df$shell_length_mm <- (latestage_df$shell_length_um)/1000</pre>
head(latestage_df)
##
       site_name site_day_key site_day_label day_trt day_an day
                     7E8E995C
## 378
          MVP-LA
                                   LA_19_Eyed 19_Eyed 15-21 19 2023-05-28
## 379
          MVP-LA
                     7E8E995C
                                   LA_19_Eyed 19_Eyed 15-21 19 2023-05-28
```

LA_19_Eyed 19_Eyed 15-21 19 2023-05-28

380

MVP-LA

7E8E995C

```
## 381
          MVP-LA
                      7E8E995C
                                   LA_19_Eyed 19_Eyed 15-21 19 2023-05-28
                     7E8E995C
## 382
          MVP-LA
                                   LA_19_Eyed 19_Eyed 15-21 19 2023-05-28
## 383
          MVP-LA
                     7E8E995C
                                   LA_19_Eyed 19_Eyed 15-21 19 2023-05-28
##
          image_label shell_length_um shell_length_mm
## 378 TX_21_small_21
                               167.430
                                              0.167430
                               316.357
## 379
        LA_19_eyed_1
                                              0.316357
## 380
         LA_19_eyed_2
                               319.598
                                              0.319598
## 381
         LA_19_eyed_3
                               332.948
                                              0.332948
## 382
         LA_19_eyed_4
                               324.306
                                              0.324306
## 383
         LA_19_eyed_5
                               332.164
                                              0.332164
#prep data for ANOVA
#use grep function to make a vector of all indices in latestage_len with the phrase "small" in them, iq
small <- grep("small", latestage_df$site_day_label, ignore.case = TRUE)</pre>
small_img <- grep("small", latestage_df$image_label, ignore.case = TRUE)</pre>
#make new dataframe WITHOUT any indices saved in small, mean remaining data are everything except small
no small df <- latestage df[-small, ]
no_small_df <- latestage_df[-small_img, ]</pre>
#rename rows
colnames(no_small_df) <- c("site_name", "site_day_key", "site_day_label", "day_trt", "day_an", "day", "</pre>
#exclude "MVP" from site names and include wild or selection line label
no_small_df["site_name"] [no_small_df["site_name"] == "MVP-TX"] <- "W1-TX"</pre>
no_small_df["site_name"] [no_small_df["site_name"] == "MVP-LA"] <- "W2-LA"</pre>
no_small_df["site_name"] [no_small_df["site_name"] == "MVP-FL"] <- "W3-FL"
no_small_df["site_name"][no_small_df["site_name"] == "MVP-LOLA"] <- "S1-LOLA"
no_small_df["site_name"] [no_small_df["site_name"] == "MVP-DEBY"] <- "S2-DEBY"</pre>
no_small_df["site_name"] [no_small_df["site_name"] == "MVP-JR"] <- "W4-VA"</pre>
no_small_df["site_name"] [no_small_df["site_name"] == "MVP-NH"] <- "W5-NH"
no_small_df["site_name"][no_small_df["site_name"] == "MVP-ME"] <- "W6-ME"
no_small_df["site_name"] [no_small_df["site_name"] == "MVP-LARMIX"] <- "H1-LARMIX"
no_small_df <- no_small_df %>%
  mutate(site_name = site_name %>%
           fct_relevel("W1-TX", "W2-LA", "W3-FL", "S1-LOLA", "S2-DEBY", "W4-VA", "W5-NH", "W6-ME", "H1-
#save latestage_df as a csv file for future analyses
write.csv(no_small_df, "../../data/performance_H2F/latestage_length.csv")
#calculate mean oyster length per site_name per day using aggregate
#make standard error equation
se <- function(x){</pre>
  sd(x)/sqrt(length(x))
}
#find mean lengths of each site_name on each day
means_mm <- aggregate(shell_length_mm ~ site_name + day_an, FUN = mean, data = no_small_df)
colnames(means_mm) <- c("site_name", "days", "mean_length_mm")</pre>
```

```
#find standard error of length of each site_name on each day
se_mm <- aggregate(shell_length_mm ~ site_name + day_an, FUN = se, data = no_small_df)</pre>
colnames(se_mm) <- c("site_name", "days", "se_length_mm")</pre>
#Check if late stage oyster data (day 19 onward, small oysters excluded) match assumptions of ANOVA
len.lm <- lm(shell_length_mm ~ day_an*site_name, data = no_small_df) #fit to lm</pre>
len.lm.resid <- len.lm$residuals #save residuals</pre>
shapiro.test(len.lm.resid) #use Shapiro Wilk test to assess normality. p <2.2e-16, data is not normal
##
##
   Shapiro-Wilk normality test
##
## data: len.lm.resid
## W = 0.9128, p-value < 2.2e-16
#plot(len.lm)
#normality is off, definite funnel in homoscedasticity plots, but KEL said to expect that what with the
#log10 transformation lm
loglen.lm <- lm(log10(shell_length_mm) ~ day_an*site_name, data = no_small_df)
loglen.lm.resid <- loglen.lm$residuals</pre>
shapiro.test(loglen.lm.resid)#p < 2.2e-16, no change in normality
##
##
   Shapiro-Wilk normality test
##
## data: loglen.lm.resid
## W = 0.88572, p-value < 2.2e-16
#plot(loglen.lm)
# I'm planning to use the log10 transformed data for the analysis, transformations helped get rid of so
#run ANOVA on log transformed length data
no_small_df$loglen <- log10(no_small_df$shell_length_mm)
len.anova3 <- Anova(lm(loglen ~ day_an*site_name, data = no_small_df), type = 3)</pre>
## Warning in printHypothesis(L, rhs, names(b)): one or more coefficients in the hypothesis include
##
        arithmetic operators in their names;
     the printed representation of the hypothesis will be omitted
##
## Warning in printHypothesis(L, rhs, names(b)): one or more coefficients in the hypothesis include
        arithmetic operators in their names;
##
     the printed representation of the hypothesis will be omitted
len.anova3
## Anova Table (Type III tests)
## Response: loglen
```

```
##
                    Sum Sq Df F value Pr(>F)
## (Intercept)
                    9.194
                            1 964.4116 <2e-16 ***
                            1 7134.8554 <2e-16 ***
## day an
                    68.017
                                  1.3858 0.1978
## site_name
                     0.106
                              8
## day_an:site_name 1.814
                              8
                                  23.7832 <2e-16 ***
## Residuals
                   16.731 1755
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
#significant interactive effect of site_name and age (p < 2e-16), but no significant effect of site_name
all_pval <- len.anova3$`Pr(>F)`
all_pval <- na.omit(all_pval)</pre>
#Interaction plot for length as a function of day*
no_small_df <- no_small_df %>%
  arrange(factor(site_name, levels = c("W1-TX", "W2-LA", "W3-FL", "S1-LOLA", "S2-DEBY", "W4-VA", "W5-NH
ggplot(no_small_df, aes(day_an, loglen, site_name = site_name))+
  #geom_point(data = means_mm, aes(days, log10(mean_length_mm), color = site_name))+
  geom_smooth(method = "lm", se = F, linewidth = .5, aes(color = site_name))+
  theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank())+
  xlab("Day")+
  ylab("Log10 Transformed Shell Length")+
  ggtitle("Shell Length Across Groups in the Nursery and Hatchery")+
```

scale_color_manual(values= c("#3e4989", "#31688e", "#26828e", "#440154", "#482878", "#1f9e89", "#35b779

'geom_smooth()' using formula = 'y ~ x'

Shell Length Across Groups in the Nursery and Hatchery

```
Pod Day Day
```

```
#convert site_name and day from chr to factor
no_small_df$site_name <- as.factor(no_small_df$site_name)
no_small_df$day_an <- as.factor(no_small_df$day_an)
str(no_small_df)</pre>
```

```
## 'data.frame':
                   1773 obs. of 11 variables:
                   : Factor w/ 9 levels "W1-TX", "W2-LA", ...: 1 1 1 1 1 1 1 1 1 1 1 ...
## $ site_name
## $ site_day_key : chr "6A2F7D94" "6A2F7D94" "6A2F7D94" "6A2F7D94" ...
## $ site_day_label : chr
                           "TX_19_Eyed" "TX_19_Eyed" "TX_19_Eyed" "TX_19_Eyed" ...
## $ day_trt
                    : chr "19_Eyed" "19_Eyed" "19_Eyed" "19_Eyed" ...
##
   $ day_an
                    : Factor w/ 2 levels "15-21", "78": 1 1 1 1 1 1 1 1 1 1 ...
## $ day
                    : int 19 19 19 19 19 19 19 19 19 ...
                    : chr "2023-05-28" "2023-05-28" "2023-05-28" "2023-05-28" ...
## $ date
                   : chr "TX_19_eyed_1" "TX_19_eyed_2" "TX_19_eyed_2" "TX_19_eyed_3" ...
## $ image_label
## $ shell_length_um: num 278 307 314 301 298 ...
## $ shell_length_mm: num 0.278 0.307 0.314 0.301 0.298 ...
                    : num -0.556 -0.512 -0.502 -0.522 -0.525 ...
##
   $ loglen
```

#subsetting based on oyster age (days) for oysters on day 78 (end of hatchery stage, pre field deployme predep <- subset(no_small_df, day_an == "78")

##ANOVA for end of hatchery (day 78) shell lengths only

```
#Linera model and ANOVA for predep shell lengths
predep$site_name <- as.factor(predep$site_name)</pre>
predep.lm <- lm(shell_length_mm ~ site_name, data = predep) #fit to lm
predep.lm.resid <- predep.lm$residuals #save residuals</pre>
shapiro.test(predep.lm.resid) #use Shapiro Wilk test to assess normality
##
## Shapiro-Wilk normality test
##
## data: predep.lm.resid
## W = 0.9974, p-value = 0.08176
# p = 0.08176, fail to reject HO, data are normal
#plots to assess if data meet assumptions of 2-way ANOVA
#plot(predep.lm)
#this data looks good, normal, homoscedastic, no outliers
#I should use Anova function to have control over the sum of squares type, but Tukey Kramer test won't
#Tukey-Kramer labels only allow one - per comparison label, so I will replace the - with _
predep$site_name <- gsub("-", "_", predep$site_name)</pre>
predep.anova3 <- Anova(lm(shell_length_mm ~ site_name, data = predep))</pre>
predep.anova3
## Anova Table (Type II tests)
##
## Response: shell_length_mm
              Sum Sq Df F value
                                     Pr(>F)
                        8 51.029 < 2.2e-16 ***
## site name 4509.8
## Residuals 11765.3 1065
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
predep.anova <- aov(lm(shell_length_mm ~ site_name, data = predep))</pre>
summary(predep.anova)
##
                 Df Sum Sq Mean Sq F value Pr(>F)
                      4510
                             563.7
                                     51.03 <2e-16 ***
## site_name
                  8
## Residuals
               1065 11765
                              11.0
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
\#extract\ p-value
predep_pval <- summary(predep.anova)[[1]][["Pr(>F)"]][1]
print(predep_pval)
```

[1] 4.998853e-70

#post-hoc comparison- Tukey-Kramer test

```
predep.tuk <- TukeyHSD(x=predep.anova,conf.level = 0.95)</pre>
#extract significant letters from TK test and store
require(multcompView)
predep_cld <- multcompLetters4(predep.anova, predep.tuk)</pre>
predep_dt <- group_by(predep, site_name) %>%
  summarise(m=mean(shell_length_mm)) %>%
  arrange(desc(m))
predep_cld <- as.data.frame.list(predep_cld$site_name)</pre>
predep_dt$predep_cld <- predep_cld$Letters</pre>
#make shell length numeric in predep
predep$shell_length_mm <- as.numeric(predep$shell_length_mm)</pre>
#P Adjust
#adjust p-value to correct for multiple tests. Make vector of p-values from each ANOVA
p_values <- c(all_pval, predep_pval)</pre>
p_values
## [1] 4.034677e-169 0.000000e+00 1.977514e-01 6.510494e-35 4.998853e-70
p_adj <- p.adjust(p_values, method = "BH")</pre>
p_adj
## [1] 1.008669e-168 0.000000e+00 1.977514e-01 8.138117e-35 8.331421e-70
predep <- predep %>%
  mutate(site_name = site_name %>%
           fct_relevel("W1_TX", "W2_LA", "W3_FL", "S1_LOLA", "S2_DEBY", "W4_VA", "W5_NH", "W6_ME", "H1_
#plot. fatten = NULL will remove the median lines in each box, allowing me to display the mean instead.
predep_box <- ggplot(data = predep, aes(site_name, shell_length_mm))+</pre>
  geom_boxplot(color = c("#3e4989", "#31688e", "#26828e", "#440154", "#482878", "#1f9e89", "#35b779", "#6
    scale_x_discrete(name = "Broodstock Group", limits = c("W1_TX", "W2_LA", "W3_FL", "S1_LOLA", "S2_DEB
  stat_summary(fun.data = mean_se, geom = "errorbar", width = 0.2, linewidth = .4, color = c("white", "
  stat_summary(fun = mean, geom = "errorbar", aes(ymax = ..y.., ymin = ..y..), width = 0.75, size = .4,
  ylab("Day 78 Shell Length (mm)")+
  xlab("Broodstock Group")+
  labs(subtitle = "p < 0.001")+
  ylim(c(0,32))+
  geom_text(data = predep_dt, size = 3, aes(label = predep_cld, y = 29, x = site_name), vjust = -0.5)+
  ggtitle("Pre-Field-Deployment (Day 78) Shell Lengths")+
  theme bw()+
  theme(plot.subtitle = element_text(size = 9), axis.title.y = element_text(size = 11), axis.title.x =
```

```
## Warning: Using 'size' aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use 'linewidth' instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```

predep_box

Warning: The dot-dot notation ('..y..') was deprecated in ggplot2 3.4.0. ## i Please use 'after_stat(y)' instead. ## This warning is displayed once every 8 hours. ## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was ## generated. ## Warning: Removed 1 row containing missing values or values outside the scale range ## ('geom_segment()'). ## Removed 1 row containing missing values or values outside the scale range ## ('geom_segment()'). ## Removed 1 row containing missing values or values outside the scale range ## ('geom segment()'). ## Removed 1 row containing missing values or values outside the scale range ## ('geom_segment()'). ## Removed 1 row containing missing values or values outside the scale range ## ('geom_segment()'). ## Removed 1 row containing missing values or values outside the scale range ## ('geom_segment()'). ## Removed 1 row containing missing values or values outside the scale range ## ('geom_segment()'). ## Removed 1 row containing missing values or values outside the scale range ## ('geom_segment()'). ## Removed 1 row containing missing values or values outside the scale range ## ('geom_segment()').



