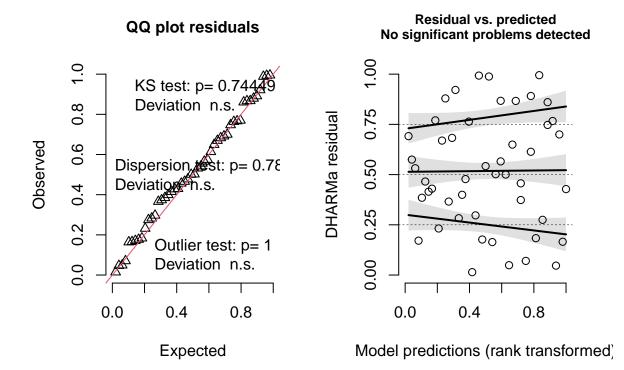
Binomial Model: Graphs and Stats

Bri Over

2024-12-11

```
# Load the dataset and alter to be used in making the model
nudi_data <- read.csv("clean_data/Meyknecht_Over_Parker_MRNE475_2024_combinedraw.csv")%>%
 mutate(temp_grp = as.factor(temp_grp)) #change how the data is structured so that treatment group is
nudi_data$trial_type[nudi_data$trial_type=="conspecific"] <- "conspecific" #ensure all cue types are l
# View the structure of the updated dataset to confirm changes:
str(nudi_data)
## 'data.frame':
                   48 obs. of 9 variables:
## $ nudi num
                    : int 2233445566...
## $ temp_grp
                   : Factor w/ 4 levels "12", "14", "16", ...: 3 3 4 4 1 1 2 2 1 1 ...
                   : chr "conspecific" "predator" "conspecific" "predator" ...
## $ trial_type
## $ mean_angle
                    : num 156 160 195 188 311 ...
## $ sd_angle
                    : num 102 112 103 105 100 ...
                    : num   0.001379   0.001311   0.001302   0.000949   0.000613   ...
## $ mean_vel
                    : num 0.00059 0.000769 0.00089 0.000788 0.000358 ...
## $ sd_vel
## $ avg_length_m : num 0.033 0.033 0.037 0.037 0.0095 0.0095 0.04 0.04 0.034 0.034 ...
## $ direction_binom: int 0 0 1 1 1 0 0 1 0 0 ...
#create the binomial model, looking at treatment group, body length and cue type as factors
#along with the interaction between treatment group and body length
glm_binom2 <- glmmTMB(direction_binom ~ temp_grp * avg_length_m + trial_type, family = binomial(link =</pre>
simulateResiduals(glm_binom2)%>%
 plot() #plot residuals to visually determine fit of model, if no red text or lines data fits the bino
```

DHARMa residual

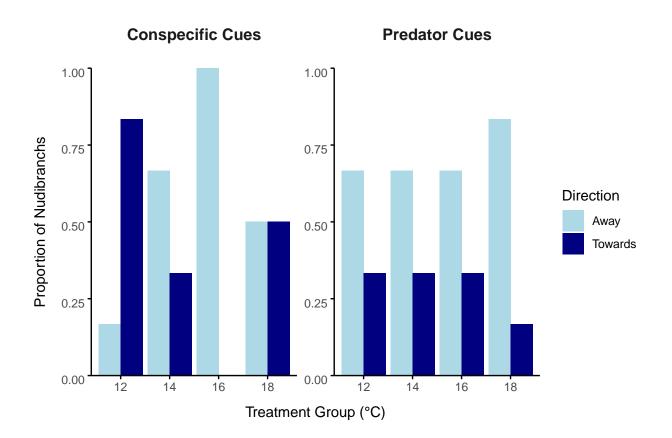


```
#Residuals show that the binomial model fits the data
# Extract the coefficients directly from the model summary
binom_table <- as.data.frame(summary(glm_binom2)$coefficients$cond)
summary(glm_binom2) #check data before exporting</pre>
```

```
Family: binomial (logit)
                      direction_binom ~ temp_grp * avg_length_m + trial_type
## Formula:
  Data: nudi_data
##
##
        AIC
                 BIC
                        logLik deviance df.resid
##
       64.2
                81.0
                         -23.1
                                   46.2
##
##
## Conditional model:
                            Estimate Std. Error z value Pr(>|z|)
##
## (Intercept)
                              1.6804
                                         1.6808
                                                   1.000
                                                           0.3174
## temp_grp14
                            -17.5703
                                        10.5224
                                                  -1.670
                                                           0.0950
                                                 -1.196
## temp_grp16
                            -13.5196
                                        11.3046
                                                           0.2317
## temp_grp18
                             -7.1760
                                         3.5701
                                                  -2.010
                                                           0.0444 *
                            -40.9577
                                                  -0.637
## avg_length_m
                                        64.3139
                                                           0.5242
## trial_typepredator
                             -0.7745
                                         0.7328
                                                  -1.057
                                                           0.2906
## temp_grp14:avg_length_m 453.1847
                                       276.1705
                                                   1.641
                                                           0.1008
## temp_grp16:avg_length_m 349.7955
                                       321.5785
                                                   1.088
                                                           0.2767
## temp_grp18:avg_length_m 205.3878
                                       116.0277
                                                           0.0767 .
                                                   1.770
```

```
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
# Save the table as a CSV file
write.csv(binom_table, "bri_code/Meyknecht_Over_Parker_MRNE475_2024_binom_table.csv", row.names = TRUE)
# Summarize the data: Count nudibranchs that moved towards or away for each combination of trial_type,
nudi_data_summary <- nudi_data %>%
  group_by(trial_type, temp_grp, direction_binom) %>%
  summarise(count = n(), .groups = 'drop')
# Calculate total number of nudibranchs in each trial_type and temp_grp
nudi_data_totals <- nudi_data %>%
  group_by(trial_type, temp_grp) %>%
  summarise(total = n(), .groups = 'drop')
# Join the summarized data with the total counts to calculate proportions
nudi data prop <- nudi data summary %>%
 left_join(nudi_data_totals, by = c("trial_type", "temp_grp")) %>%
  mutate(proportion = count / total) # Calculate proportion for each direction
# Ensure 'direction_binom' is a factor with appropriate levels
nudi_data_prop$direction_binom <- factor(nudi_data_prop$direction_binom, levels = c(0, 1))</pre>
# Handle missing data for "towards" direction in case there are no nudibranchs for that direction
nudi_data_prop_complete <- nudi_data_prop %>%
  complete(trial_type, temp_grp, direction_binom, fill = list(count = 0)) # Ensures missing rows are a
# Set any NA values in proportion to O (this handles cases where count is O or missing)
nudi_data_prop_complete$proportion[is.na(nudi_data_prop_complete$proportion)] <- 0</pre>
# Get the maximum proportion to set the y-axis limit
y_max <- max(nudi_data_prop_complete$proportion, na.rm = TRUE)</pre>
# Create the bar chart with the correct facet to have both conspecific and predator cues, with the same
plot_rxn_probs <- ggplot(nudi_data_prop_complete, aes(x = factor(temp_grp), y = proportion, fill = fact</pre>
  geom_bar(stat = "identity", position = "dodge") + # dodge bars so they do not overlap
  facet_wrap(~ trial_type, scales = "free_y",
             labeller = labeller(trial_type = c("conspecific" = "Conspecific Cues", "predator" = "Predator"
  labs(
   title = "", #Set axes and legend title and remove graph title
   x = "Treatment Group (°C)",
   y = "Proportion of Nudibranchs",
   fill = "Direction"
  scale_fill_manual(values = c("0" = "lightblue", "1" = "navyblue"), labels = c("Away", "Towards")) +
  scale_y\_continuous(limits = c(0, y\_max), expand = c(0, 0)) + #Proportion bars touch the x-axis
  scale_x_discrete(expand = c(0.2, 0)) + \# Increase space between the treatment groups on the x-axis
  theme minimal() +
  theme(
   axis.text.y = element_text(vjust = 1), # Adjust vertical alignment of y-axis text
   panel.grid.major = element_blank(),
   panel.grid.minor = element_blank(),
   axis.line = element_line(color = "black", linewidth = 0.5),
```

```
axis.ticks = element_line(color = "black", linewidth = 0.5),
axis.title.x = element_text(margin = margin(t = 10)), # Add space between x-axis labels and number
axis.title.y = element_text(margin = margin(r = 10)), # Add space between y-axis labels and number
strip.text = element_text(size = 12, face = "bold", margin = margin(b = 20)) # Add space between f
)
plot_rxn_probs
```



#optional look at summary table for proportion values of bar chart, may save and include
#double check that each treatment group is represented properly

summary_table <- nudi_data_prop_complete %>%
 dplyr::select(trial_type, temp_grp, direction_binom, proportion) %>%
 arrange(trial_type, temp_grp, direction_binom)

print(summary_table)

```
## # A tibble: 16 x 4
##
      trial_type temp_grp direction_binom proportion
##
      <chr>
                  <fct>
                            <fct>
                                                  <dbl>
    1 conspecific 12
                                                  0.167
##
                            Ω
##
    2 conspecific 12
                            1
                                                  0.833
                            0
                                                  0.667
##
   3 conspecific 14
   4 conspecific 14
                            1
                                                  0.333
   5 conspecific 16
                            0
```

```
## 6 conspecific 16
                                              0
                          1
## 7 conspecific 18
                          0
                                              0.5
## 8 conspecific 18
                                              0.5
## 9 predator
                 12
                          0
                                              0.667
## 10 predator
                 12
                                              0.333
                          1
## 11 predator
               14
                          0
                                              0.667
## 12 predator
                14
                                              0.333
                          1
## 13 predator
                 16
                          0
                                              0.667
## 14 predator
                 16
                          1
                                              0.333
## 15 predator
                 18
                          0
                                              0.833
## 16 predator
                 18
                          1
                                              0.167
```

```
#save plot to git folder
```

ggsave(plot = plot_rxn_probs, filename = here("bri_code", "Meyknecht_Over_Parker_MRNE475_2024_plot_rxn_