YSP 2022_Aggression and social behavior under risk

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In Summer 2022, our two YSP students Sophia and Ishita tested whether Amazon mollies altered their 1) aggressive and 2) sociable behavior when under the threat of predation risk. They designed two assays to measure these behaviors and then repeatedly measured individual fish in either or both contexts (with risk, or without risk).

Aggressive behavior: Study measures bites as count data for the duration of an 5 minute trial Absence of risk (N = 10), Presence of risk (N = 10) were tested 2x each for their specific treatment (testing interval = 8 days). No individual experienced both treatments. 1 individual per treatment died (Total N = 18, or N = 9 per treatment)

Sociability: Study measures association time with conspecific 3N for the duration of an 15 minute trial. $N = 10 \ 3N$ mollies. All fish experienced both treatments in a balanced design. Each fish was tested in control and presence settings (3 trials each, for a total of 6 trials per fish). No deaths.

Load data

```
library(lme4)
## Loading required package: Matrix
library(lmerTest)
## Attaching package: 'lmerTest'
## The following object is masked from 'package:lme4':
##
##
       lmer
## The following object is masked from 'package:stats':
##
##
       step
library(tidyverse)
## -- Attaching core tidyverse packages -----
                                                    ----- tidyverse 2.0.0 --
## v dplyr
               1.1.4
                         v readr
                                     2.1.5
```

```
## v forcats
               1.0.0
                                     1.5.1
                         v stringr
               3.5.0
## v ggplot2
                         v tibble
                                     3.2.1
## v lubridate 1.9.3
                         v tidyr
                                     1.3.1
## v purrr
               1.0.2
## -- Conflicts -----
                                       ------tidyverse_conflicts() --
## x tidyr::expand() masks Matrix::expand()
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                     masks stats::lag()
## x tidyr::pack() masks Matrix::pack()
## x tidyr::unpack() masks Matrix::unpack()
## 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(patchwork)
library(ggplot2)
library(dplyr)
library(ggpubr)
library(MuMIn)
#Kate's computer
setwd("C:/Users/katel/Box Sync/YSP data")
agg.data <- read.csv("Aguinaga et al_ Risk response data_for deposit.csv")
social.data <- read.csv("Aguinaga et al_Sociability data_for deposit.csv")</pre>
head(agg.data)
         ID trial bites
## 1 1 GGGG
                     27 Predator absent
                1
## 2 2 GGGG
                     41 Predator absent
## 3 3 GYYY
                     29 Predator absent
                1
## 4 4 GYYY
                2
                     84 Predator absent
## 5 5 GGYP
                1
                     96 Predator absent
## 6 6 GGYP
                     O Predator absent
head(social.data)
         ID trial total time
## 1 1 GYPP
                1
                    587.512 Predator absent
## 2 2 GYPP
                2
                     322.198 Predator absent
## 3 3 GYPP
                3
                     300.258 Predator absent
## 4 4 GYPP
                1 734.612 Predator present
                2
## 5 5 GYPP
                     782.018 Predator present
## 6 6 GYPP
                     596.894 Predator present
```

Data cleaning

Jon provided the data in a few different files. Here I clean up the data to compile each behavioral assay into one data file. This code no longer needs to be run as the cleaned files are loaded up above.

```
dataAbsence <- read.csv("IshitaYSPAbsenceCSV.csv")</pre>
data <- dataAbsence %>%
  select(ID, T1A, T2A) %>%
  pivot_longer(!ID, names_to = "trial", values_to = "bites") %>%
  mutate(trial = replace(trial, trial == "T1A", 1),
         trial = replace(trial, trial == "T2A", 2),
         context = "Predator absent")
dataPresence <- read.csv("IshitaYSPPresenceCSV.csv")</pre>
data2 <- dataPresence %>%
  pivot_longer(!ID, names_to = "trial", values_to = "bites") %>%
  mutate(trial = replace(trial, trial == "T1P", 1),
         trial = replace(trial, trial == "T2P", 2),
         context = "Predator present")
risk.behav <- rbind(data, data2)
write.csv(risk.behav, file = "Risk response YSP.csv")
####
sophia <- read.csv("SophiaYSPCSV.csv")</pre>
sophia2 <- sophia %>%
pivot_longer(!ID, names_to = "trial", values_to = "total_time") %>%
 mutate(context = ifelse(trial == "T1A" | trial == "T2A" | trial == "T3A", "Predator absent", "Predator
  mutate(trial = replace(trial, trial == "T1A" | trial == "T4P", 1),
         trial = replace(trial, trial == "T2A" | trial == "T5P", 2),
         trial = replace(trial, trial == "T3A" | trial == "T6P", 3))
write.csv(sophia2, file = "sociability YSP.csv")
```

Aggression data

Testing to see whether number of bites differs based on the context (pred present, pred absent) or across trials.

```
mod1.agg <- lmer(bites ~ trial *context + (1|ID), data = agg.data)

## boundary (singular) fit: see help('isSingular')

ranova(mod1.agg) # no evidence of individual differences here

## ANOVA-like table for random-effects: Single term deletions

## Model:

## bites ~ trial + context + (1 | ID) + trial:context

## npar logLik AIC LRT Df Pr(>Chisq)

## <none> 6 -178.54 369.08

## (1 | ID) 5 -178.54 367.08 5.6843e-14 1 1
```

```
anova(mod1.agg) # no evidence of strong differences across context, or trial
```

```
## Type III Analysis of Variance Table with Satterthwaite's method
                 Sum Sq Mean Sq NumDF DenDF F value Pr(>F)
##
## trial
                 521.36 521.36
                                    1
                                         32 0.1670 0.6855
                 295.21 295.21
                                         32 0.0946 0.7604
## context
                                    1
## trial:context 318.03 318.03
                                         32 0.1019 0.7517
                                    1
summary(mod1.agg)
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: bites ~ trial * context + (1 | ID)
##
      Data: agg.data
##
## REML criterion at convergence: 357.1
##
## Scaled residuals:
##
      Min
               1Q Median
                                3Q
                                       Max
## -1.3364 -0.8114 0.2029 0.6071 1.9032
##
## Random effects:
                         Variance Std.Dev.
  Groups
             Name
             (Intercept)
                            0
                                   0.00
                         3122
                                  55.87
## Residual
## Number of obs: 36, groups: ID, 18
##
## Fixed effects:
##
                                 Estimate Std. Error
                                                          df t value Pr(>|t|)
## (Intercept)
                                   65.667
                                              41.644 32.000
                                                               1.577
                                                                        0.125
## trial
                                    1.667
                                              26.338 32.000
                                                               0.063
                                                                        0.950
                                              58.893 32.000 -0.308
## contextPredator present
                                  -18.111
                                                                        0.760
## trial:contextPredator present
                                   11.889
                                              37.247 32.000
                                                              0.319
                                                                        0.752
##
## Correlation of Fixed Effects:
##
               (Intr) trial cntxPp
               -0.949
## trial
## cntxtPrdtrp -0.707 0.671
## trl:cntxtPp 0.671 -0.707 -0.949
## optimizer (nloptwrap) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')
```

Because we have such a small sample size though and we didn't necessary expect there to be a strong interaction between context and trial, let's remove that interaction and re-run the model

```
mod2.agg <- lmer(bites ~ trial + context + (1|ID), data = agg.data)
## boundary (singular) fit: see help('isSingular')</pre>
```

```
ranova(mod2.agg)
## ANOVA-like table for random-effects: Single term deletions
##
## Model:
## bites ~ trial + context + (1 | ID)
           npar logLik
                          AIC LRT Df Pr(>Chisq)
            5 -183.12 376.24
## <none>
## (1 | ID)
              4 -183.12 374.24
                               0 1
anova(mod2.agg)
## Type III Analysis of Variance Table with Satterthwaite's method
          Sum Sq Mean Sq NumDF DenDF F value Pr(>F)
## trial
          521.36 521.36
                                 33 0.1717 0.6813
                          1
## context 0.69
                    0.69
                                  33 0.0002 0.9880
summary(mod2.agg)
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: bites ~ trial + context + (1 | ID)
     Data: agg.data
##
## REML criterion at convergence: 366.2
##
## Scaled residuals:
##
      Min
             1Q Median
                               3Q
                                      Max
## -1.3061 -0.8766 0.2057 0.6694 1.9836
##
## Random effects:
## Groups Name
                        Variance Std.Dev.
                                 0.00
            (Intercept)
                          0
## Residual
                        3037
                                 55.11
## Number of obs: 36, groups: ID, 18
##
## Fixed effects:
                                                  df t value Pr(>|t|)
##
                          Estimate Std. Error
## (Intercept)
                          56.7500
                                     30.4605 33.0000 1.863 0.0714 .
## trial
                            7.6111
                                      18.3684 33.0000 0.414
                                                               0.6813
## contextPredator present -0.2778
                                     18.3684 33.0000 -0.015
                                                              0.9880
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Correlation of Fixed Effects:
              (Intr) trial
              -0.905
## trial
## cntxtPrdtrp -0.302 0.000
## optimizer (nloptwrap) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')
```

```
r.squaredGLMM(mod2.agg)

## Warning: 'r.squaredGLMM' now calculates a revised statistic. See the help page.

## R2m R2c
```

Getting lots of singular fit warnings because ID accounts for no variance so will re-run without ID results still the same - no strong difference in bites across trials or contexts

Sociability data

[1,] 0.004888064 0.004888064

Now let's compare whether total time spent associating with a conspecific differed across contexts or trials

```
mod1.soc <- lmer(total_time ~ trial * context + (1|ID), data = social.data)</pre>
ranova(mod1.soc) # significant effect of individual
## ANOVA-like table for random-effects: Single term deletions
##
## Model:
## total_time ~ trial + context + (1 | ID) + trial:context
           npar logLik
                            AIC
                                   LRT Df Pr(>Chisq)
## <none>
              6 -367.80 747.61
## (1 | ID)
              5 -370.69 751.38 5.7678 1
                                             0.01632 *
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
anova(mod1.soc)
```

```
## Type III Analysis of Variance Table with Satterthwaite's method
                Sum Sq Mean Sq NumDF DenDF F value Pr(>F)
## trial
                 105829 105829
                                   1
                                        47 5.3510 0.02513 *
                  5157
                          5157
                                   1
                                        47
                                           0.2608 0.61199
## context
## trial:context
                  7393
                          7393
                                   1
                                        47
                                           0.3738 0.54389
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
```

summary(mod1.soc)

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: total_time ~ trial * context + (1 | ID)
##
      Data: social.data
##
## REML criterion at convergence: 735.6
##
## Scaled residuals:
##
      Min
                1Q Median
                                3Q
                                       Max
  -2.3288 -0.5248 0.1425 0.7145 1.9128
##
##
## Random effects:
                         Variance Std.Dev.
##
   Groups
            Name
##
  ID
             (Intercept) 6633
                                   81.45
                                  140.63
## Residual
                         19777
## Number of obs: 60, groups: ID, 10
## Fixed effects:
##
                                 Estimate Std. Error
                                                         df t value Pr(>|t|)
## (Intercept)
                                               72.65 55.69
                                                              8.442 1.52e-11 ***
                                   613.34
## trial
                                   -65.03
                                               31.45 47.00
                                                             -2.068
                                                                       0.0442 *
## contextPredator present
                                    49.06
                                               96.07 47.00
                                                              0.511
                                                                       0.6120
## trial:contextPredator present
                                    27.19
                                               44.47 47.00
                                                              0.611
                                                                       0.5439
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##
               (Intr) trial cntxPp
## trial
               -0.866
## cntxtPrdtrp -0.661 0.655
## trl:cntxtPp 0.612 -0.707 -0.926
```

Again because our sample size is so small and we didn't really expect a strong interaction, let's remove the interaction and re-run the model

We see a significant effect of individual and context - fish spend more time associating with their conspecific when a predator is present (which is exactly we expect it). We should report the parameter estimates, se and t-values from the summary and then the F-stats and p-values from the anova()

```
mod2.soc <- lmer(total_time ~ trial + context + (1|ID), data = social.data)
ranova(mod2.soc)</pre>
```

```
anova(mod2.soc)
## Type III Analysis of Variance Table with Satterthwaite's method
          Sum Sq Mean Sq NumDF DenDF F value Pr(>F)
## trial
          105829 105829
                           1
                                  48 5.4218 0.024146 *
## context 160486 160486
                             1
                                  48 8.2220 0.006131 **
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
summary(mod2.soc)
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: total_time ~ trial + context + (1 | ID)
     Data: social.data
##
## REML criterion at convergence: 745.4
## Scaled residuals:
##
              1Q Median
      Min
                               3Q
                                      Max
## -2.3383 -0.6264 0.1412 0.6503 1.9219
##
## Random effects:
                        Variance Std.Dev.
## Groups
           Name
             (Intercept) 6676
                                  81.71
## Residual
                        19519
                                 139.71
## Number of obs: 60, groups: ID, 10
##
## Fixed effects:
                          Estimate Std. Error
##
                                                  df t value Pr(>|t|)
## (Intercept)
                            586.15 57.19 49.15 10.250 8.51e-14 ***
## trial
                            -51.44
                                        22.09 48.00 -2.328 0.02415 *
                                       36.07 48.00 2.867 0.00613 **
## contextPredator present 103.44
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Correlation of Fixed Effects:
               (Intr) trial
              -0.773
## trial
## cntxtPrdtrp -0.315 0.000
r.squaredGLMM(mod2.soc)
                       R2c
             R<sub>2</sub>m
## [1,] 0.1469848 0.3643898
We can also estimate the repeatability of individual - about 0.25
```

summary(mod2.soc)

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: total time ~ trial + context + (1 | ID)
##
      Data: social.data
##
## REML criterion at convergence: 745.4
##
## Scaled residuals:
##
      Min
                1Q Median
                                30
                                       Max
## -2.3383 -0.6264 0.1412 0.6503
                                   1.9219
##
## Random effects:
##
   Groups
                         Variance Std.Dev.
                          6676
                                   81.71
##
   ID
             (Intercept)
                         19519
                                  139.71
   Residual
## Number of obs: 60, groups: ID, 10
##
## Fixed effects:
                                                   df t value Pr(>|t|)
##
                           Estimate Std. Error
## (Intercept)
                             586.15
                                         57.19
                                                49.15
                                                      10.250 8.51e-14 ***
## trial
                             -51.44
                                         22.09 48.00
                                                       -2.328 0.02415 *
                             103.44
                                                        2.867 0.00613 **
## contextPredator present
                                         36.07 48.00
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##
               (Intr) trial
## trial
               -0.773
## cntxtPrdtrp -0.315 0.000
6676/(6676+19519)
```

[1] 0.2548578

While the interaction between trial and context is not significant, the graph makes it appear that there could be a slight trend for greater habituation in the control fish, compared to the predator cue fish. As such, we'll perform a post hoc to look at this in more detail. We recognize that these results should be interpreted very cautiously and are more just descriptive of the patterns we're seeing rather than a proper test.

```
social.control <- social.data %>%
   filter(context == "Predator absent")

social.pred <- social.data %>%
   filter(context == "Predator present")

control1 <- lmer(total_time ~trial + (1|ID), data = social.control)
anova(control1)

## Type III Analysis of Variance Table with Satterthwaite's method
## Sum Sq Mean Sq NumDF DenDF F value Pr(>F)
## trial 84581 84581 1 19 4.2765 0.05253 .
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

```
summary(control1)
```

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: total_time ~ trial + (1 | ID)
     Data: social.control
##
## REML criterion at convergence: 368
## Scaled residuals:
       Min
                10
                     Median
                                   30
## -2.53853 -0.45990 -0.01597 0.41489 1.99979
##
## Random effects:
## Groups Name
                        Variance Std.Dev.
## ID
            (Intercept) 5069
                                  71.2
                                 140.6
## Residual
                        19778
## Number of obs: 30, groups: ID, 10
## Fixed effects:
              Estimate Std. Error
                                   df t value Pr(>|t|)
##
## (Intercept) 613.34
                            71.57 26.92
                                         8.570 3.57e-09 ***
## trial
                -65.03
                            31.45 19.00 -2.068 0.0525 .
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Correlation of Fixed Effects:
        (Intr)
## trial -0.879
pred1 <- lmer(total_time ~ trial + (1|ID), data = social.pred)</pre>
anova(pred1)
## Type III Analysis of Variance Table with Satterthwaite's method
        Sum Sq Mean Sq NumDF DenDF F value Pr(>F)
## trial 28640
                 28640
                                19 1.3487 0.2599
                          1
summary(pred1)
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: total_time ~ trial + (1 | ID)
##
     Data: social.pred
## REML criterion at convergence: 370.8
##
## Scaled residuals:
           1Q Median
      Min
                               3Q
                                      Max
## -1.9624 -0.4979 0.1934 0.8109 1.1883
## Random effects:
```

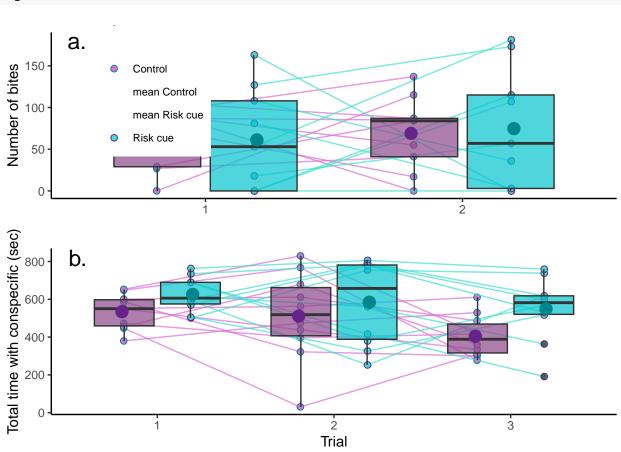
```
## Groups
                        Variance Std.Dev.
            Name
             (Intercept) 6684
## TD
                                  81.76
## Residual
                        21236
                                 145.73
## Number of obs: 30, groups: ID, 10
## Fixed effects:
              Estimate Std. Error
                                      df t value Pr(>|t|)
                            74.99 27.26
## (Intercept)
                662.40
                                          8.833 1.74e-09 ***
## trial
                -37.84
                            32.59 19.00 -1.161
                                                     0.26
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Correlation of Fixed Effects:
        (Intr)
##
## trial -0.869
```

Figures

```
agg.data <- read.csv("Aguinaga et al_ Risk response data_for deposit.csv")
social.data <- read.csv("Aguinaga et al_Sociability data_for deposit.csv")</pre>
agg.data$trial <- factor(agg.data$trial)</pre>
social.data$trial <- factor(social.data$trial)</pre>
agg.data <- agg.data%>%
 mutate(context = recode(context,
                          "Predator absent" = "Control",
                          "Predator present" = "Risk cue"))
social.data <- social.data%>%
  mutate(context = recode(context,
                          "Predator absent" = "Control",
                          "Predator present" = "Risk cue"))
agg.plot <- ggplot(agg.data, aes(x = trial, y = bites, fill = context)) +
  geom_line(data = agg.data[agg.data$context == "Control",],
            aes(x = trial, y = bites, group = ID, color = context),
            position = position nudge(x = -0.19),
            linewidth = 0.5, color = "orchid", alpha = 0.7) +
  geom_point(data = agg.data[agg.data$context =="Control",],
             aes(x = trial, y = bites, group = ID, color = context),
             position = position_nudge(x = -0.19),
             size = 2, shape = 21, fill = "orchid") +
  geom_line(data = agg.data[agg.data$context =="Risk cue",],
            aes(x = trial, y = bites, group = ID, color = context),
            position = position_nudge(x = 0.19),
            linewidth = 0.5, color = "turquoise", alpha = 0.7) +
  geom_point(data = agg.data[agg.data$context == "Risk cue",],
             aes(x = trial, y = bites, group = ID, color = context),
             position = position_nudge(x = 0.19),
             size = 2, shape = 21, fill = "turquoise") +
```

```
geom_boxplot(data = agg.data, aes(x = trial, y = bites, fill = context), alpha = 0.7) +
  scale_fill_manual(values = c("orchid4", "turquoise3")) +
  stat_summary(fun.y = "mean", geom = "point", aes(color=paste("mean", context)), position = position_d
  scale_color_manual(values = c("turquoise4", "darkorchid4", "turquoise4", "darkorchid4")) +
  ylab("Number of bites") +
  xlab("") +
  theme_classic() +
  theme(legend.title = element_blank(),
       legend.position = c(0.18, 0.8),
       legend.text = element_text(size = 8)) +
  annotate("text", label = "a.", x = 0.5, y = 180, size = 6)
## Warning: The 'fun.y' argument of 'stat_summary()' is deprecated as of ggplot2 3.3.0.
## i Please use the 'fun' argument instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
## Warning: A numeric 'legend.position' argument in 'theme()' was deprecated in ggplot2
## 3.5.0.
## i Please use the 'legend.position.inside' argument of 'theme()' instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
soc.plot <- ggplot(social.data, aes(x = trial, y = total_time, fill = context)) +</pre>
  geom_line(data = social.data[social.data$context == "Control",],
            aes(x = trial, y = total_time, group = ID, color = context),
            position = position_nudge(x = -0.19),
            linewidth = 0.5, color = "orchid", alpha = 0.7) +
  geom_point(data = social.data[social.data$context =="Control",],
             aes(x = trial, y = total_time, group = ID, color = context),
             position = position_nudge(x = -0.19),
             size = 2, shape = 21, fill = "orchid") +
  geom line(data = social.data[social.data$context =="Risk cue",],
            aes(x = trial, y = total_time, group = ID, color = context),
            position = position_nudge(x = 0.19),
            linewidth = 0.5, color = "turquoise", alpha = 0.7) +
  geom_point(data = social.data[social.data$context == "Risk cue",],
             aes(x = trial, y = total_time, group = ID, color = context),
             position = position_nudge(x = 0.19),
             size = 2, shape = 21, fill = "turquoise") +
  geom_boxplot(data = social.data, aes(x = trial, y = total_time, fill = context), alpha = 0.7) +
  scale_fill_manual(values = c("orchid4", "turquoise3")) +
  stat_summary(fun.y = "mean", geom = "point", aes(color=paste("mean", context)), position = position_d
  scale_color_manual(values = c("turquoise4", "darkorchid4", "turquoise4", "darkorchid4")) +
  ylab("Total time with conspecific (sec)") +
  xlab("Trial") +
  theme_classic() +
  theme(legend.title = element blank(),
        legend.position = "none") +
    annotate("text", label = "b.", x = 0.55, y = 820, size = 6)
```

```
fig <- ggpubr::ggarrange(agg.plot, soc.plot, ncol = 1)
fig</pre>
```



ggsave(fig, file = "YSP figures.pdf", height = 180, width = 120, units = "mm", dpi = 600)
#NB: I removed one of the legend keys in Adobe after the fact as I could not figure it out in R!