

Guppy Network Infection Experiment

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1 Introduction

1.1 Overall summary

This is an analysis to examine how host traits, particularly host sociality and sex, impacts the spread of disease within experimental epidemics of Trinidadian guppies and Monogenean parasites.

1.2 Quick experimental summary

We released 6 female guppies and 3 male guppies into experimental behavioral enclosures together and recorded their baseline sociality (Here defined as Contact Rate) for 2 days. We then introduced an infected female index fish on day 3 to see how quickly parasites spread within the group of fish and how likely specific individuals to pick up parasite infection. We repeated recording behavior for 7 days post infection for a total of 9 days of behavioral recordings, however the analysis within this document is limited to days 1-3.

1.3 Questions

1.3.1 Individual level

- Are females more likely to be infected than male hosts?
- Does host worm contact rate explain individual level probability of infection?
- Do infected index fish receive less contact rate from uninfected conspecifics relative to uninfected index fish?

-Do infected index fish with higher infection intensity receive less contact rate from uninfected conspecifics?

1.4 Data description, structure and type

This analysis uses a single master dataframe titled: `DyadindividualContacts_20240726.csv` that is subset into different portions based on what questions/analysis we are addressing. Variables within the data frame as follows:

Ind: The individual ID of each fish used in the trial.

ContactRate: The daily average contact rate for each individual (*contacts/dyad/min*)

Day3ContactRatewoI: The daily average contact rate for each non-index individual with other non-index individuals. Index fish is the fish introduced on Day 3 with infection (or not for uninfected controls) (*contacts/dyad/min*)

IndexContact: The daily average contact rate with the index individual. Repeated for each individual since they only had one day with the index individual. (*contacts/dyad/min*)

Day: The day in which the behavior was taken. Days 1 and 2 are days without the infected index individual and day three is with the infected index individual. 3 levels. One level for each day.

Popavg: The average contact rate for each group (*Contacts/dyads/min/fish*)

Pop: The group ID for each replicate. One level per group.

Treatment: Whether the group received an infected or uninfected index fish at the beginning of day 3. 2 levels.

TempTreatment: The treatment (infected or uninfected) of each group for each day. For every group it will be Unf on days 1 and 2 due to not having an index, but Inf for Infected treatments for days 3. 2 levels.

Sex: The sex of each individual. 2 levels, F for female, M for male.

Inf.: This is renamed to Infection in some subset data sets. Whether or not an individual was infected at the end of day 3. 2 levels, 1 - infected, 0-uninfected

worms: The number of worms on each individual at the end of day 3.

prev: Prevalence (proportion of infected individuals) for the group at the end of day 3.

IndexWorm: Number of worms on the index fish at the end of day 3. This is used as a metric for the amount of worms an individual will have been exposed to during initial epidemic spread.

Index: Whether the individual was an index fish or not. 2 levels, 0-not index, 1-index

ChConWithIn: Change in contact with including the contacts with the index between days 2 and 3.

ChConWOIn: Change in contact with with only non-index individuals between days 2 and 3.

ChConPop: Change in the group average contact rate between days 2 and 3

PreSVL: The snout to vent length of the guppy prior to introduction into behavioral arena. Measured in (*Millimeters*)

PreWeight: The weight of the individual prior to introduction into the behavior box. Measured in (*grams*)

PostSVL: The snout to vent length of the guppy after wrapping up entire experiment. Measured in (*Millimeters*)

PostWeight: The Weight of the guppy after wrapping up entire experiment. Measured in (*grams*)

```

# Load in libraries for analysis and data wrangling Visualization
library(ggplot2)
library(visreg)
source("http://highstat.com/Books/BGS/GAMM/RCodeP2/HighstatLibV6.R")
# (generalized) Linear mixed modeling
library(lme4)
library(glmmTMB)
# Statistical analysis reporting and model validation
library(performance)
library(car)
library(lmtest)
library(DHARMA)
# Data wrangling
library(dplyr)
library(plyr)
library(tidyverse)
library(emmeans)

# Importing the master data sheet check intro for description of
# factors Choose X datasheet (specify when finished)
IndContacts <- read.csv("DyadindividualContacts_20240726.csv")

# This is creating a variable to calculate the contact rate with the
# infected individual for non-infected fish and combining them back
# together. Subsetting to individuals who are not the index
# Additionally this code subsets down to only focusing on dyads with
# the index fish as the recipient

IndContactsNI <- IndContacts %>%
  filter(Index != "1" & day == "3" & str_detect(recipID2, "_10")) %>%
  dplyr::rename(IndexContactR = ContactInitR) %>%
  select(c(fishID2, recipID2, IndexContactR))

# Subsetting to individuals who are the index fish
IndContactsI <- IndContacts %>%
  filter(Index == "1") %>%
  mutate(IndexContactR = NA) %>%
  select(c(fishID2, recipID2, IndexContactR))
# Rbinding our two variables together
IndContactsNI <- rbind(IndContactsNI, IndContactsI)

# Merging our dataframe back together by fishID to add in the
# indexcontactrate
IndContacts2 <- merge(IndContacts, IndContactsNI, by = c("fishID2"), .keep_all = TRUE) %>%
  # select(-c(recipID2.y, day.y, population.y))%>%
  dplyr::rename(recipID2 = recipID2.x) %>%
  distinct(fishID2, recipID2, population, day, .keep_all = TRUE)

# Sum contact rate across all videos in a day for each dyad We want
# to combine all contacts across a day. We added 0.001 to each of
# these to help with further model fit without impacting the
# distribution of the data

```

```

IndContactsSum <- IndContacts2 %>%
  select(-c(X, fishID, frame_num, recipID)) %>%
  mutate(ContactInitR = ContactInitR + 1e-04, TotalCR = TotalCR + 1e-04,
         RecipCR = RecipCR + 1e-04) %>%
  group_by(fishID2, recipID2, day) %>%
  dplyr::summarise(ContactInitRs = sum(ContactInitR), TotalCRs = sum(TotalCR),
                  RecipCRs = sum(RecipCR), contactinit = sum(contactinit), RecipfledCR = sum(RecipfledCR))

# Remove unneeded columns
IndContactssub <- IndContacts2 %>%
  select(-c(X, fishID, frame_num, recipID, ContactInitR, TotalCR, RecipCR,
            RecipfledCR)) %>%
  group_by(fishID2, day)

# Merge sheets together so that new data frame has all variables
IndContacts3 <- merge(IndContactsSum, IndContactssub, by = c("fishID2",
                  "recipID2", "day"), .keep_all = TRUE)

# Going down to one row per individual
IndContacts3 <- IndContacts3 %>%
  group_by(fishID2, recipID2, population, day) %>%
  distinct(fishID2, recipID2, population, day, .keep_all = TRUE) %>%
  filter(fishID2 != "RCinf1_NA")

# Calculating important metrics that may apply across all subsetting
# of data

# Wormcontacts = Number of contacts a fish has had with infectious
# agents. So the multiplicative of contact rate with the infected
# index with the number of worms on the infected index
IndContacts3 <- IndContacts3 %>%
  mutate(WormContact = IndexContactR * IndexWorm)

# Assigning a fish label to the fish so that if we want to subset
# down to one metric for fish because not all metrics change across
# days. This currently doesnt work due to index only have day 3
# measure Fish<-c('1','0','0')

# Adding this factor to overall data frame
# PopulationContacts$Fish<-as.factor(Fish)

# Overall summary just checking that everything looks okay.
summary(IndContacts3)

```

```

##      fishID2      recipID2      day      ContactInitRs
## Length:3213      Length:3213      Min.   :1.000      Min.   :0.000100
## Class :character  Class :character  1st Qu.:1.000      1st Qu.:0.005731
## Mode  :character  Mode  :character  Median :2.000      Median :0.015294
##                                     Mean  :2.098      Mean  :0.023234
##                                     3rd Qu.:3.000      3rd Qu.:0.031961
##                                     Max.   :3.000      Max.   :0.210114
##
##      TotalCRs      RecipCRs      contactinit      RecipfledCR
## Min.   :0.00010      Min.   :0.00010      Min.   : 0.0      Min.   :0.00000

```

```
## 1st Qu.:0.02560 1st Qu.:0.00585 1st Qu.: 195.0 1st Qu.:0.00450
## Median :0.06421 Median :0.01518 Median : 516.0 Median :0.01226
## Mean :0.09574 Mean :0.02325 Mean : 766.5 Mean :0.01860
## 3rd Qu.:0.13188 3rd Qu.:0.03188 3rd Qu.:1060.0 3rd Qu.:0.02548
## Max. :0.76783 Max. :0.21011 Max. :7558.0 Max. :0.19029
##
## population TotalContacts contactinit recipinit
## Length:3213 Min. : 0 Min. : 0.0 Min. : 0.0
## Class :character 1st Qu.: 897 1st Qu.: 195.0 1st Qu.: 202.0
## Mode :character Median : 2192 Median : 516.0 Median : 515.0
## Mean : 3178 Mean : 766.5 Mean : 767.1
## 3rd Qu.: 4405 3rd Qu.:1060.0 3rd Qu.:1061.0
## Max. :27629 Max. :7558.0 Max. :7558.0
##
## recipfled Sex recipsex worms
## Min. : 0.0 Length:3213 Length:3213 Min. : 0.00
## 1st Qu.: 154.0 Class :character Class :character 1st Qu.: 0.00
## Median : 416.0 Mode :character Mode :character Median : 0.00
## Mean : 615.7 Mean : 10.16
## 3rd Qu.: 852.0 3rd Qu.: 1.00
## Max. :5987.0 Max. :214.00
## NA's :2313
## Recipworm Index Infectionstat InfectionTrt
## Min. : 0.00 Min. :0.00000 Min. :0.00000 Min. :0.0000
## 1st Qu.: 0.00 1st Qu.:0.00000 1st Qu.:0.00000 1st Qu.:0.0000
## Median : 0.00 Median :0.00000 Median :0.00000 Median :1.0000
## Mean : 10.16 Mean :0.03922 Mean :0.09804 Mean :0.7087
## 3rd Qu.: 1.00 3rd Qu.:0.00000 3rd Qu.:0.00000 3rd Qu.:1.0000
## Max. :214.00 Max. :1.00000 Max. :1.00000 Max. :1.0000
## NA's :2313
## IndexWorm recipID2.y IndexContactR WormContact
## Min. : 32.00 Length:3213 Min. :0.00000 Min. : 0.0000
## 1st Qu.: 43.00 Class :character 1st Qu.:0.00336 1st Qu.: 0.2600
## Median : 79.00 Mode :character Median :0.00948 Median : 0.8368
## Mean : 94.05 Mean :0.01789 Mean : 2.1099
## 3rd Qu.:114.00 3rd Qu.:0.02156 3rd Qu.: 2.4295
## Max. :214.00 Max. :0.18141 Max. :14.3317
## NA's :936 NA's :126 NA's :1026
```

```
# Set data strcuture for some factors
```

```
# Setting Day as a factor instead of a numeric
```

```
IndContacts3$day <- as.factor(IndContacts3$day)
```

```
# Setting whether a fish was an index (intitial infection) or not as factor
```

```
IndContacts3$Index <- as.factor(IndContacts3$Index)
```

```
# Setting whether an fish was infected or not as factor
```

```
IndContacts3$Infectionstat <- as.factor(IndContacts3$Infectionstat)
```

```
# Setting whether fish were male or female as a factor
```

```
IndContacts3$Sex <- as.factor(IndContacts3$Sex)
```

```
IndContacts3$recipsex <- as.factor(IndContacts3$recipsex)
```

```
# Setting the treatment of fish group (infected index or not) as factor
```

```
IndContacts3$InfectionTrt <- as.factor(IndContacts3$InfectionTrt)
```

```

# Setting population the fish were from as a factor
IndContacts3$population <- as.factor(IndContacts3$population)
# Setting fish ID as a factor
IndContacts3$fishID2 <- as.factor(IndContacts3$fishID2)
# Setting recipient ID as a factor
IndContacts3$recipID2 <- as.factor(IndContacts3$recipID2)

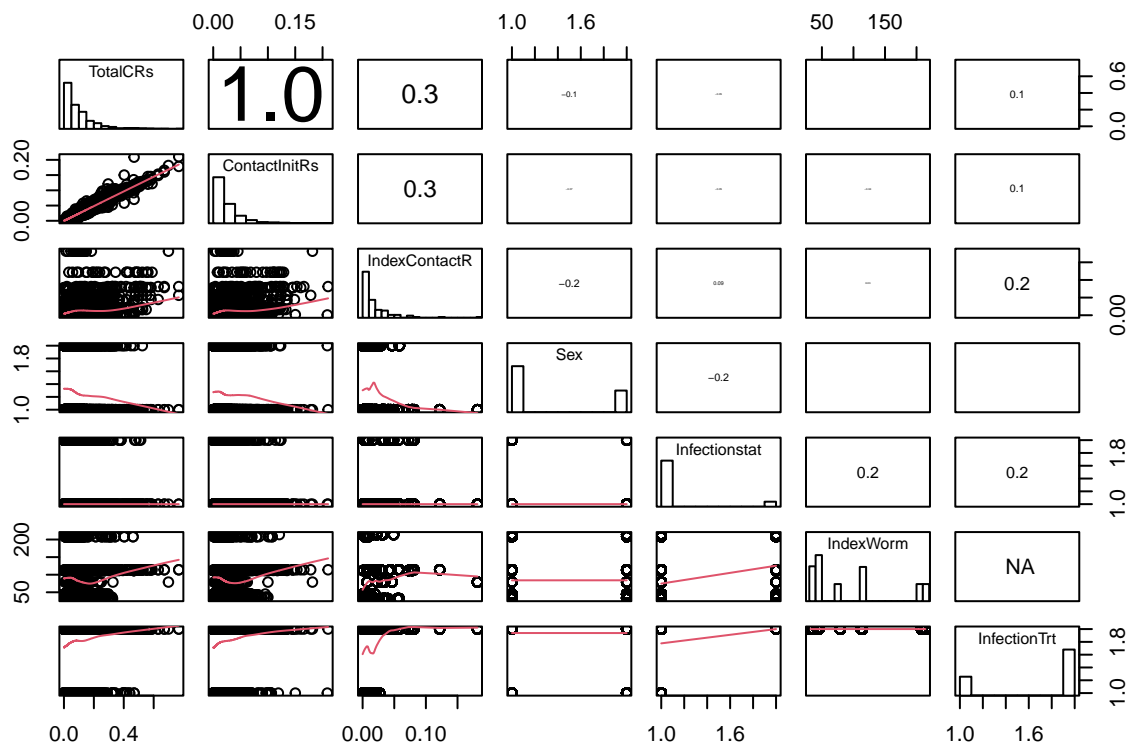
```

1.5 Initial data exploration

```

# Pairs plot to look at autocorrelation and potential relationships
# before diving into analysis
pairs(~TotalCRs + ContactInitRs + IndexContactR + Sex + Infectionstat +
      IndexWorm + InfectionTrt, lower.panel = panel.smooth, diag.panel = panel.hist,
      upper.panel = panel.cor, data = IndContacts3)

```



```

# Checking to see if there are any day differences
daylm <- glmmTMB(TotalCRs ~ day + (1 | population), family = beta_family(),
                 IndContacts3)
# summary of the model
summary(daylm)

```

```

## Family: beta ( logit )
## Formula:      TotalCRs ~ day + (1 | population)
## Data: IndContacts3
##
##      AIC      BIC   logLik deviance df.resid
## -10194.6 -10164.3   5102.3 -10204.6     3208

```

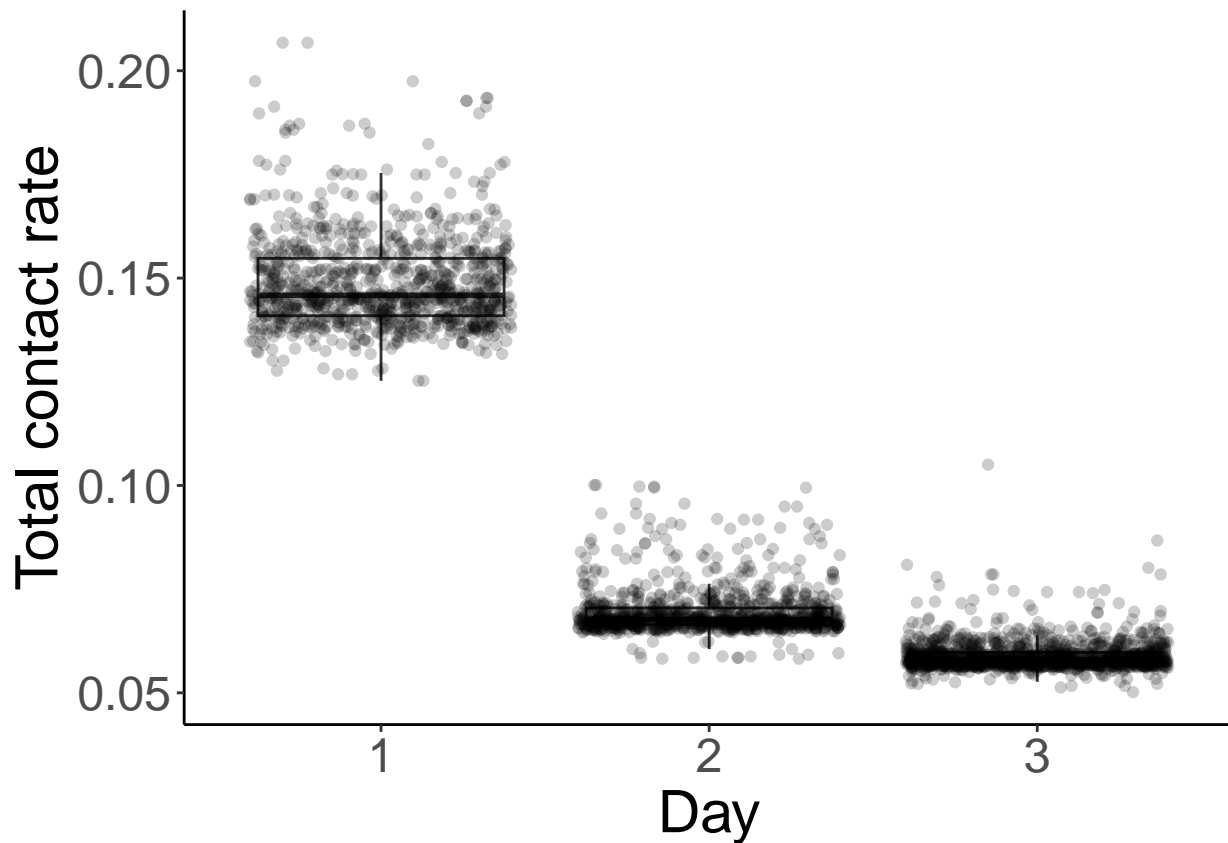
```
##
## Random effects:
##
## Conditional model:
##   Groups      Name      Variance Std.Dev.
## population (Intercept) 0.1611  0.4014
## Number of obs: 3213, groups: population, 14
##
## Dispersion parameter for beta family (): 16
##
## Conditional model:
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.71784    0.10950  -15.69  <2e-16 ***
## day2        -0.85657    0.03289  -26.04  <2e-16 ***
## day3        -1.02701    0.03251  -31.59  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# Testing for significance
drop1(daylm, test = "Chisq")
```

```
## Single term deletions
##
## Model:
## TotalCRs ~ day + (1 | population)
##      Df      AIC    LRT Pr(>Chi)
## <none>   -10194.6
## day     2  -9273.4 925.29 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# It seems we have some high contact rate on day 1, we have to
# eliminate it due to acclimation
```

```
# Plotting behavior by individual across groups
daylmVR <- visreg(daylm, "day", scale = "response", partial = TRUE, plot = FALSE)
# Generate the fitted relationship
daylmVRfit <- daylmVR$fit
# Generate the residuals
daylmVRres <- daylmVR$res
# Plot this relationship
ggplot(daylmVRres, aes(x = day, y = visregRes)) + geom_boxplot(outliers = FALSE) +
  geom_jitter(alpha = 0.2) + theme_classic() + theme(legend.position = "none",
    text = element_text(size = 22)) + ylab("Total contact rate") + xlab("Day")
```



```
# There are significant differences in day so we have to subset out
# first day We need to remove first day measurements.
IndContactPW <- IndContacts3 %>%
  filter(day == "2" | day == "3" & Index == "0" & recipID2 != str_detect(recipID2,
    "_10")) %>%
  select(fishID2, day, ContactInitRs) %>%
  pivot_wider(names_from = day, values_from = ContactInitRs) %>%
  na.omit(2) %>%
  mutate(ChContactinit = `3` - `2`) %>%
  select(fishID2, recipID2, ChContactinit)
```

```
## Adding missing grouping variables: 'recipID2', 'population'
## Adding missing grouping variables: 'population'
```

```
# Create a index only data frame and set its change in contact rate
# as NA given it only came in on day 3
IndContactI <- IndContacts3 %>%
  filter(Index == "1") %>%
  mutate(ChContactinit = NA)

IndContacts4 <- merge(IndContactPW, IndContacts3, by = c("fishID2", "recipID2",
  "population")) %>%
  filter(day != "1")

IndContacts4 <- rbind(IndContacts4, IndContactI)
```



```
# %>% filter(population == 'ADinf1' | population == 'ADunf1' |
# population == 'GZinf1' | population == 'GZunf1' | population ==
# 'SBinf1' | population == 'SBunf1')
```

2 Individual Level Analysis

2.1 Are hosts with higher worm contact rate more likely to become infected?

For this first bit we need to select the data we need to answer this specific question

```
# Filtering down to only infected treatments, individuals who are not
# the index, one measure per fish and days 2 and 3. Note the days 2
# and 3 is because initial viewing of behaviors across days looks
# like fish were not behaving normally on day 1 due to acclimation to
# the new environment.
IndContactsPI1 <- IndContacts3 %>%
  filter(Index != "1" & day == "3" & InfectionTrt == "1" & Sex == "F" &
    str_detect(recipID2, "_10"))
```

2.1.1 Fitting the contact rate versus probability of infection model

We're fitting a binomial generalized linear mixed model with whether an individual was infected or not at the end of day 3 as our response variable and worm contact rate as predictor variables. We have population as a random effect to control for non-independence of worm contact rate between individuals within the same group.

In addition to this, we fit a second GLMM examining the same relationship but with host contact rate instead of the product term of worm contact rate to see if it better fits our data.

```
# generalized linear mixed model to determine if individual sociality
# explained the probability of becoming infected
ProbinfLMF <- glmmTMB(Infectionstat ~ WormContact + I(WormContact^2) +
  (1 | population), family = binomial(), data = IndContactsPI1)

# We fit a second model with just contact rate to see if it does a
# better job than fitting the product term of contact rate and
# infection intensity of the index A model with just social contacts
# to test against using the social contacts
ProbinfLMFSC <- glmmTMB(Infectionstat ~ ContactInitRs + I(ContactInitRs^2) +
  (1 | population), family = binomial(), data = IndContactsPI1)

# Summary of our two models
summary(ProbinfLMF)
```

```
## Family: binomial ( logit )
## Formula:
## Infectionstat ~ WormContact + I(WormContact^2) + (1 | population)
## Data: IndContactsPI1
```

```
##
##      AIC      BIC    logLik deviance df.resid
##      72.9     81.3     -32.4     64.9      56
##
## Random effects:
##
## Conditional model:
## Groups      Name          Variance Std.Dev.
## population (Intercept) 0.4255   0.6523
## Number of obs: 60, groups: population, 10
##
## Conditional model:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -1.75787    0.57614  -3.051  0.00228 **
## WormContact      1.03939    0.37157   2.797  0.00515 **
## I(WormContact^2) -0.07959    0.03400  -2.341  0.01925 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

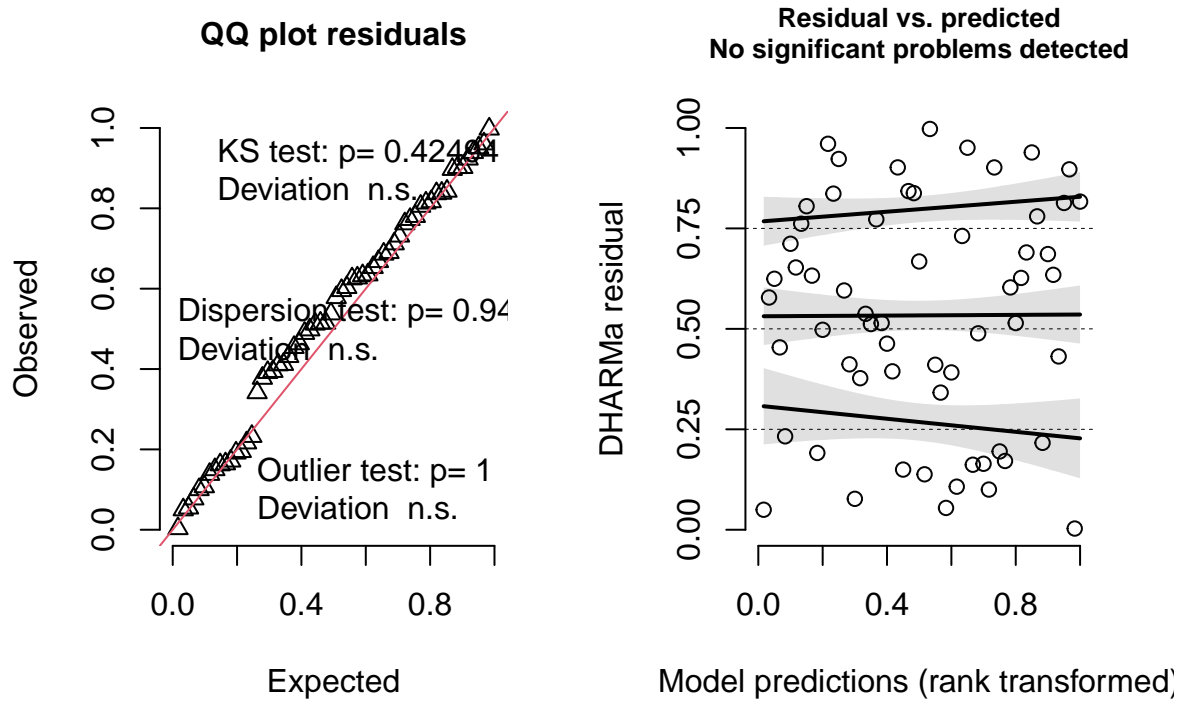
```
summary(ProbinfLMFSC)
```

```
## Family: binomial ( logit )
## Formula:
## Infectionstat ~ ContactInitRs + I(ContactInitRs^2) + (1 | population)
## Data: IndContactsPI1
##
##      AIC      BIC    logLik deviance df.resid
##      78.3     86.7     -35.1     70.3      56
##
## Random effects:
##
## Conditional model:
## Groups      Name          Variance Std.Dev.
## population (Intercept) 1.702    1.305
## Number of obs: 60, groups: population, 10
##
## Conditional model:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -1.4958    0.8086  -1.850  0.0643 .
## ContactInitRs     69.5154    44.7783   1.552  0.1206
## I(ContactInitRs^2) -615.9132   453.3311  -1.359  0.1743
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

2.1.2 Next we see will validate the model to make sure everything is looking good.

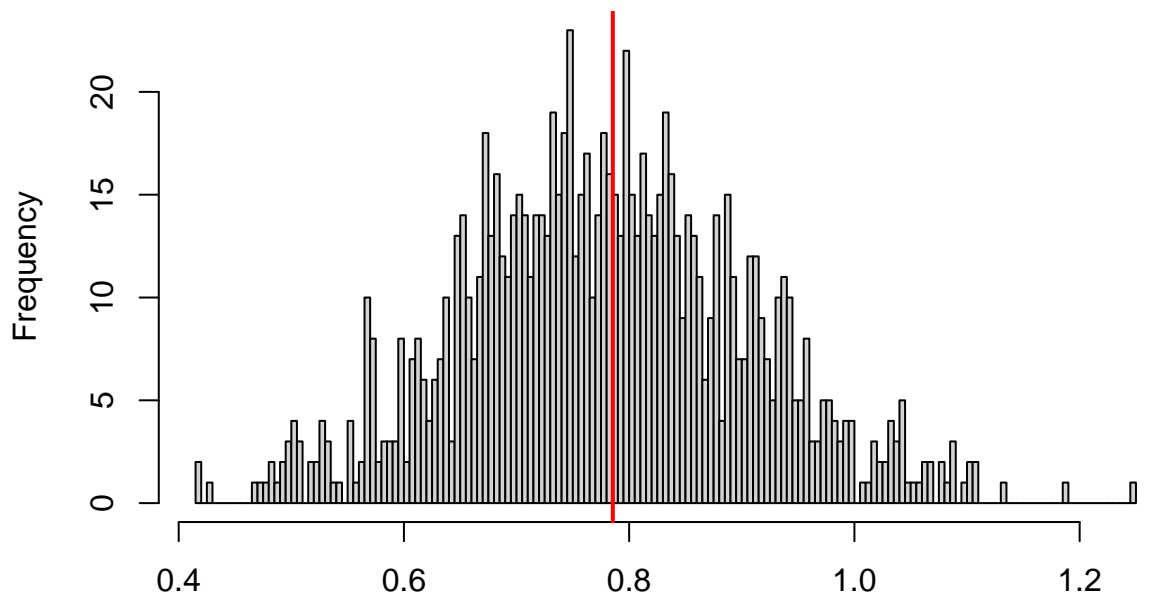
```
# Using DHARMa package to look at quantile residuals to validate the
# worm contact rate model fit
sim_residuals_ProbinfLM <- simulateResiduals(ProbinfLMF, 1000)
# Quantile residuals Plotting the quantile residuals
plot(sim_residuals_ProbinfLM)
```

DHARMa residual



```
# Testing dispersion
testDispersion(sim_residuais_ProbinfLM)
```

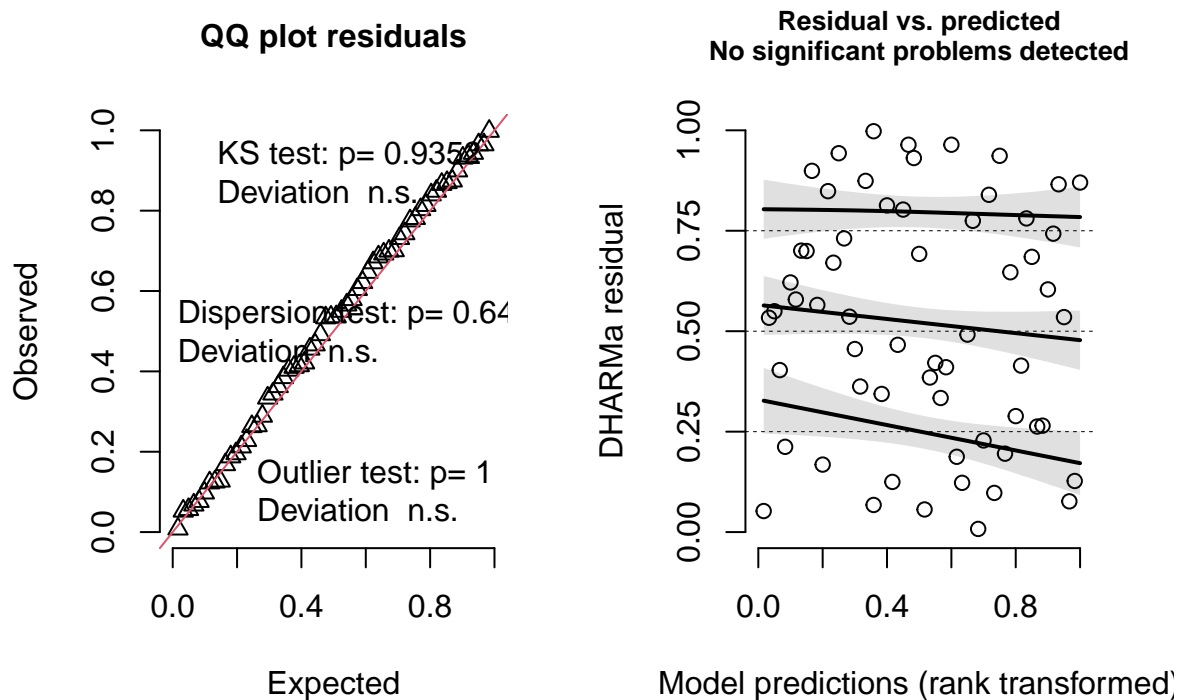
DHARMa nonparametric dispersion test via sd of residuals fitted vs. simulated



```
##
## DHARMA nonparametric dispersion test via sd of residuals fitted vs.
## simulated
##
## data: simulationOutput
## dispersion = 1.0087, p-value = 0.942
## alternative hypothesis: two.sided
```

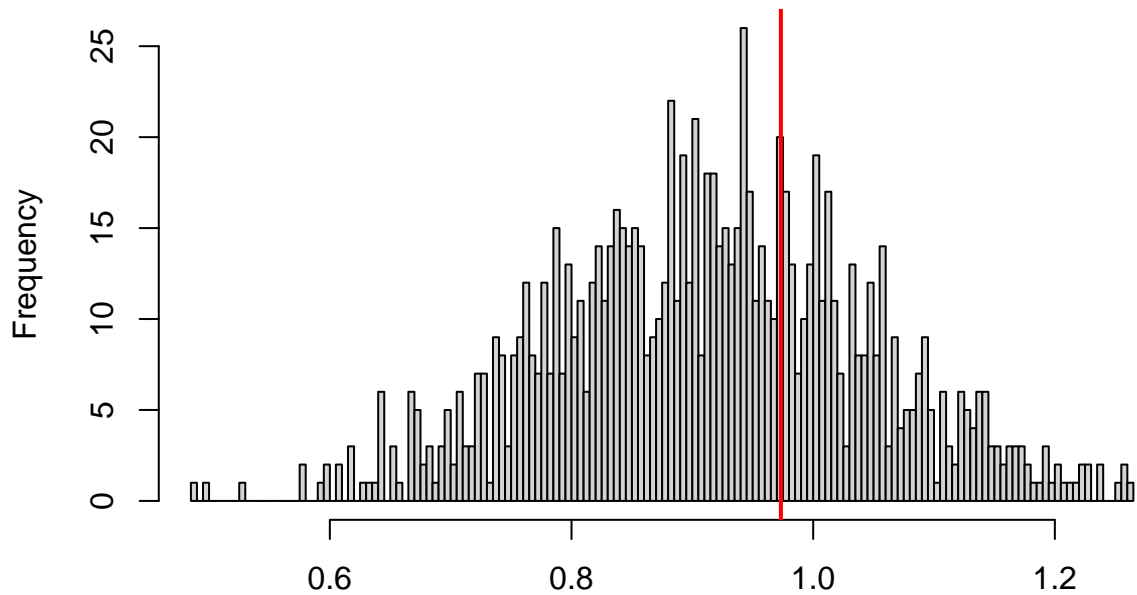
```
# Using DHARMA package to look at quantile residuals to validate the
# contact rate model fit
sim_residuals_ProbinfLMSC <- simulateResiduals(ProbinfLMFSC, 1000)
# Quantile residuals Plotting the quantile residuals
plot(sim_residuals_ProbinfLMSC)
```

DHARMA residual



```
# Testing dispersion
testDispersion(sim_residuals_ProbinfLMSC)
```

DHARMa nonparametric dispersion test via sd of residuals fitted vs. simulated



Simulated values, red line = fitted model. p-value (two.sided) = 0.648

```
##
## DHARMa nonparametric dispersion test via sd of residuals fitted vs.
## simulated
##
## data: simulationOutput
## dispersion = 1.065, p-value = 0.648
## alternative hypothesis: two.sided
```

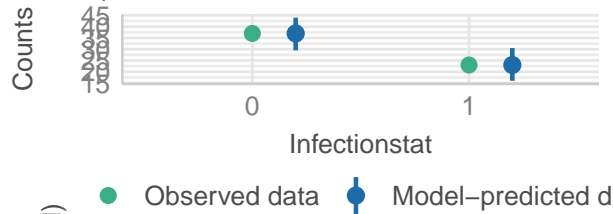
```
# Below we are examining the quantile residuals among the fixed
# variables to see if there are any funky patterns. Excluded from
# main run for space but remove # to see plots
# plotResiduals(sim_residuals_InfCRLm, IndividualInfDFD2$Sex)
# plotResiduals(sim_residuals_InfCRLm, IndividualInfDFD2$ContactRate)
```

```
check_model(ProbinfLMF)
```

```
## 'check_outliers()' does not yet support models of class 'glmmTMB'.
```

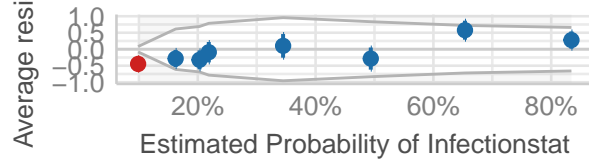
Posterior Predictive Check

Model-predicted intervals should include observed



Binned Residuals

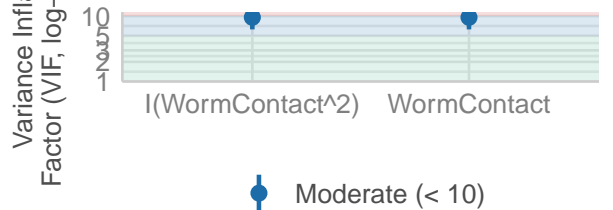
Points should be within error bounds



Within error bounds ● no ● yes

Collinearity

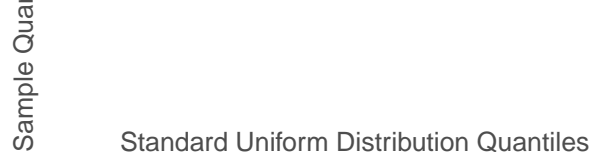
High collinearity (VIF) may inflate parameter uncertainty



● Moderate (< 10)

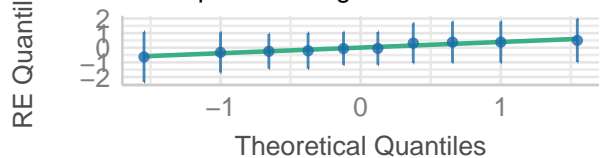
Uniformity of Residuals

Points should fall along the line



Normality of Random Effects (population)

Points should be plotted along the line



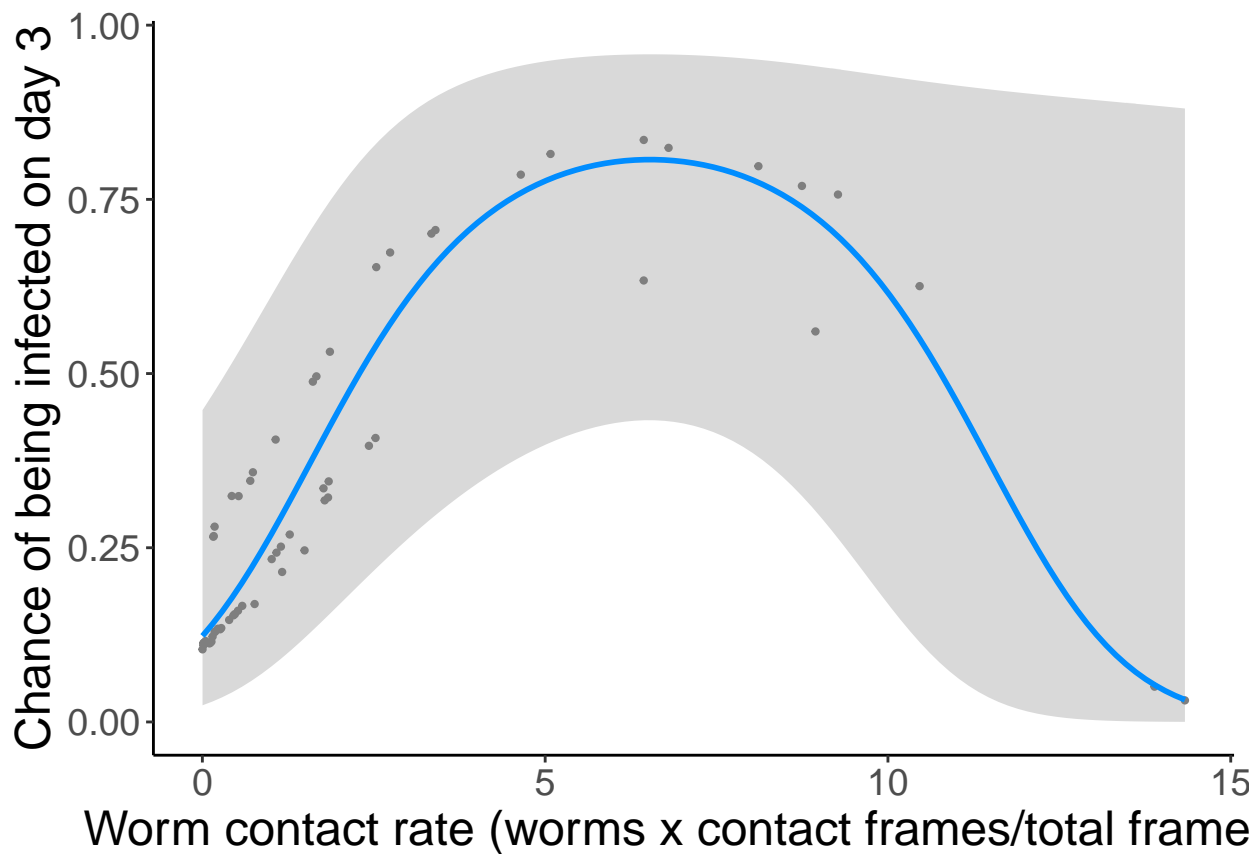
2.1.3 Likelihood ratio test to test for significance in our model

```
# Drop 1 with chisq test to do a likelihood ratio test for our model
drop1(ProbinfLMF, test = "Chisq")
```

```
## Single term deletions
##
## Model:
## Infectionstat ~ WormContact + I(WormContact^2) + (1 | population)
##           Df    AIC    LRT Pr(>Chi)
## <none>          72.889
## WormContact      1 80.081  9.1926 0.002430 **
## I(WormContact^2)  1 79.620  8.7310 0.003128 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

2.1.4 Visualization of the relationships between worm contact rate and probability of infection

```
# Visreg of contact rate effect
InfWCRvr <- visreg(ProbinfLMF, "WormContact", scale = "response", partial = T,
  overlay = TRUE, gg = TRUE) + xlab("Worm contact rate (worms x contact frames/total frames) ") +
  ylab("Chance of being infected on day 3") + theme_classic() + theme(text = element_text(size = 18))
print(InfWCRvr)
```



2.2 Do hosts contact uninfected index as much as infected index individuals?

```
# Filtering down to only Getting index contact rate
IndContactsIndexrecip <- IndContacts4 %>%
  filter(Index == "1")
```

2.2.1 Fitting a model to test for total contact rate for infected versus uninfected index individuals

```
# Logging and scaling contact rate due to heavy skew and the desire
# to be comparable between explanatory variables
IndContactsIndexrecip$ScContactInitRs <- scale(log10(IndContactsIndexrecip$ContactInitRs))

# Fitting the glmmTMB for received contacts by infection treatment
# of the group
IndexTC1m <- glmmTMB(RecipCRs ~ InfectionTrt + recipsex + ScContactInitRs +
  I(ScContactInitRs^2) + (1 | fishID2), family = beta_family(link = "logit"),
  data = IndContactsIndexrecip)

# summary of the model
summary(IndexTC1m)
```

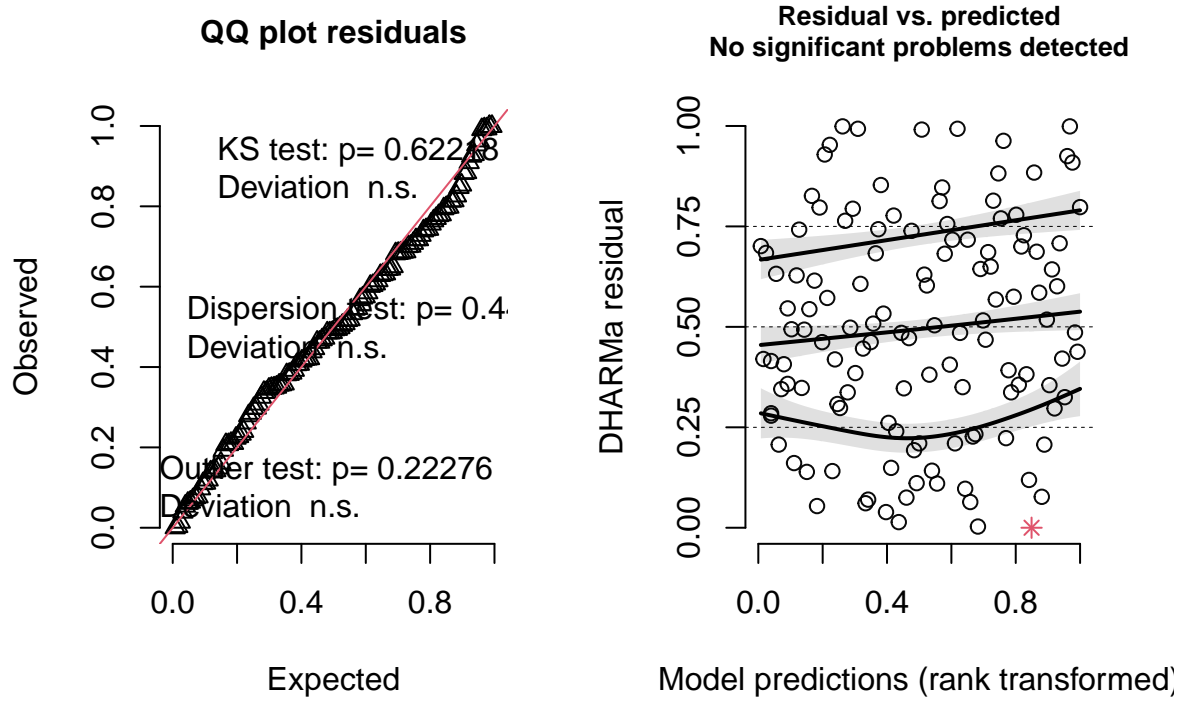
```
## Family: beta ( logit )
```

```
## Formula:
## RecipCRs ~ InfectionTrt + recipsex + ScContactInitRs + I(ScContactInitRs^2) +
## (1 | fishID2)
## Data: IndContactsIndexrecip
##
##      AIC      BIC    logLik deviance df.resid
## -1025.0 -1005.2    519.5  -1039.0      119
##
## Random effects:
##
## Conditional model:
## Groups Name      Variance Std.Dev.
## fishID2 (Intercept) 0.01133  0.1064
## Number of obs: 126, groups: fishID2, 14
##
## Dispersion parameter for beta family (): 478
##
## Conditional model:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -4.82983    0.10340  -46.71 < 2e-16 ***
## InfectionTrt1   -0.06979    0.11463   -0.61 0.542613
## recipsexM       0.34583    0.08221    4.21 2.59e-05 ***
## ScContactInitRs  1.30765    0.06359   20.56 < 2e-16 ***
## I(ScContactInitRs^2) 0.14887    0.03936    3.78 0.000155 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

2.2.2 Validating the recipient initiated contact with index model

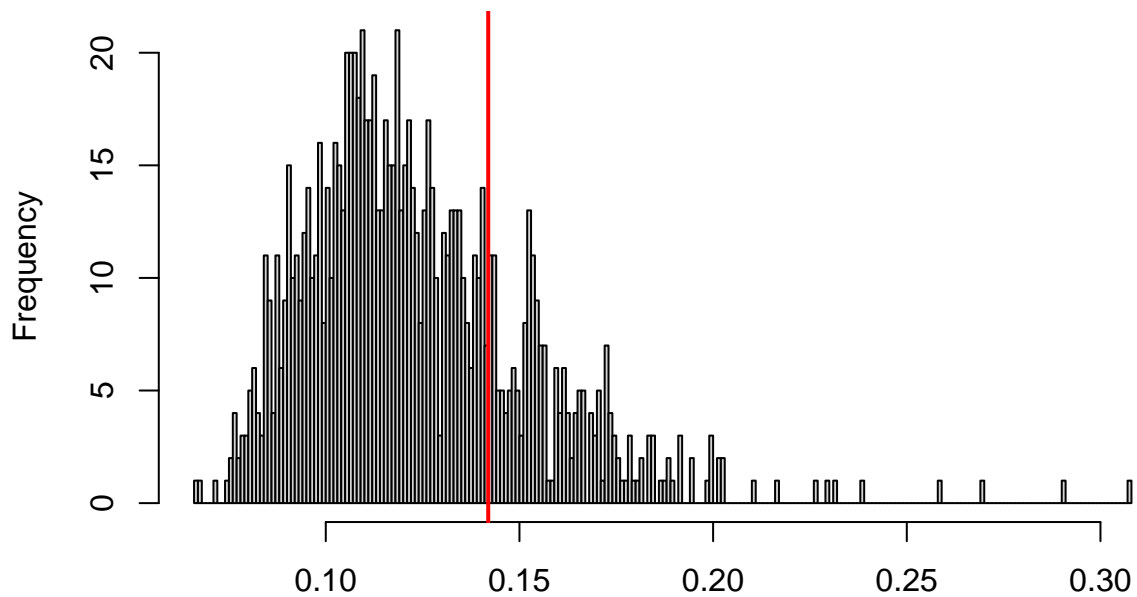
```
sim_residuals_IndexTC1m <- simulateResiduals(IndexTC1m, 1000) #Quantile residuals
# Plotting the quantile residuals
plot(sim_residuals_IndexTC1m)
```


DHARMa residual



```
# Testing dispersion
testDispersion(sim_residuals_IndexTC1m)
```

DHARMa nonparametric dispersion test via sd of residuals fitted vs. simulated



Simulated values, red line = fitted model. p-value (two.sided) = 0.44

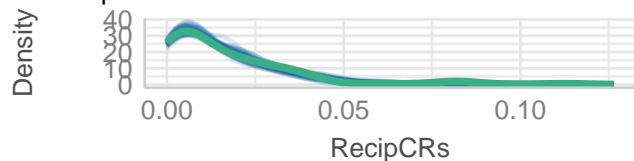
```
##
## DHARMA nonparametric dispersion test via sd of residuals fitted vs.
## simulated
##
## data: simulationOutput
## dispersion = 1.1523, p-value = 0.44
## alternative hypothesis: two.sided
```

```
# using the checkmodel function to validate model
check_model(IndexTC1m)
```

```
## 'check_outliers()' does not yet support models of class 'glmmTMB'.
```

Posterior Predictive Check

Model-predicted lines should resemble observed data



Homogeneity of Variance

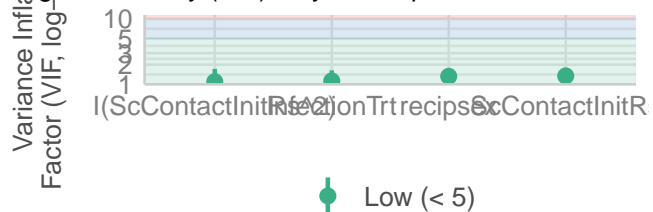
Reference line should be flat and horizontal



— Observed data — Model-predicted data

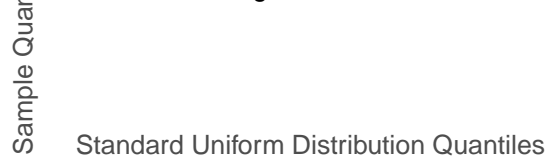
Collinearity

High collinearity (VIF) may inflate parameter uncertainty



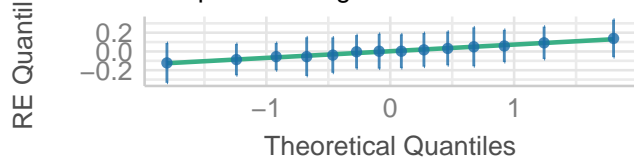
Uniformity of Residuals

Dots should fall along the line



Normality of Random Effects (fishID2)

Dots should be plotted along the line



```
# sim_residuals_IndexInitTC1m <-simulateResiduals(IndexInitTC1m,
# 1000) #Quantile residuals #Plotting the quantile residuals
# plot(sim_residuals_IndexInitTC1m) #Testing dispersion
# testDispersion(sim_residuals_IndexInitTC1m)
```

2.2.3 Using a likelihood ratio test to test for significance in the recipient initiated contact with index model

```
drop1(IndexTC1m, test = "Chisq")
```

```
## Single term deletions
```

```
##
## Model:
## RecipCRs ~ InfectionTrt + recipsex + ScContactInitRs + I(ScContactInitRs^2) +
## (1 | fishID2)
##           Df          AIC      LRT Pr(>Chi)
## <none>          -1025.04
## InfectionTrt      1 -1026.68   0.359 0.5491749
## recipsex          1 -1011.25  15.794 7.063e-05 ***
## ScContactInitRs    1  -844.04 183.003 < 2.2e-16 ***
## I(ScContactInitRs^2) 1 -1013.99  13.054 0.0003027 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

2.2.4 Visualizing the relationship between recipient sex and total contacts with index

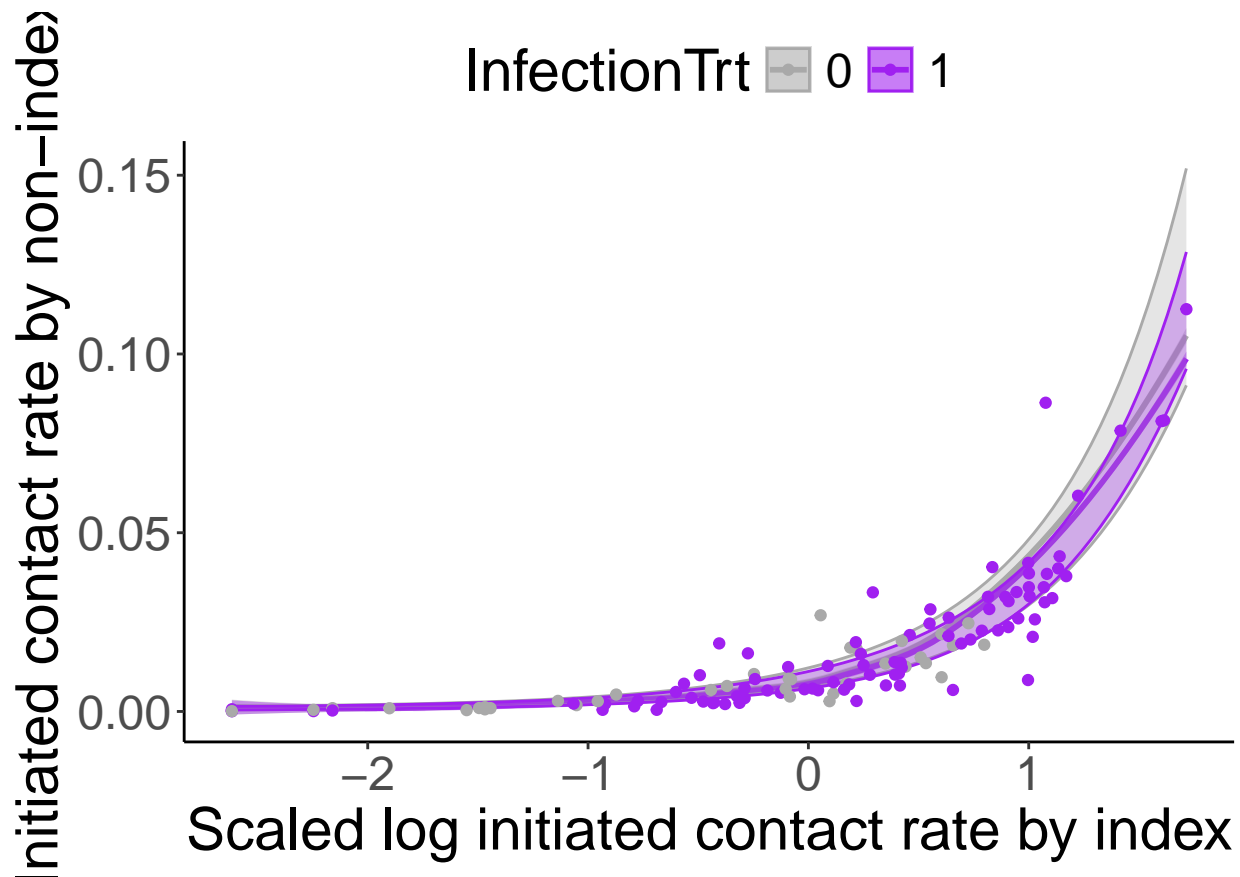
```
# Generating fitted relationship and residuals for contact rate and
# infection treatment
RCind <- visreg(IndexTCLm, "ScContactInitRs", "InfectionTrt", scale = "response",
  partial = T, plot = FALSE)

# Grabbing the fitted relationship
RCindfit <- RCind$fit
# Grabbing the partial residuals
RCinfres <- RCind$res
# Setting the color scheme for infected and uninfected groups
cpinf = c("darkgray", "purple")

# Plotting the relationship for received contact rate and infection
# treatment

ggplot(RCindfit, aes(x = ScContactInitRs, y = visregFit, group = InfectionTrt,
  colour = InfectionTrt, fill = InfectionTrt)) + geom_smooth(data = RCindfit,
  method = "loess", aes(x = ScContactInitRs, y = visregFit)) + geom_ribbon(aes(ymin = visregLwr,
  ymax = visregUpr), alpha = 0.3) + geom_point(data = IndContactsIndexrecip,
  aes(x = ScContactInitRs, y = RecipCRs)) + theme_classic() + xlab("Scaled log initiated contact rate")
  ylab("Initiated contact rate by non-index") + theme(text = element_text(size = 22)) +
  theme(legend.position = "top") + scale_color_manual(values = cpinf) +
  scale_fill_manual(values = cpinf)

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



2.3 Do index fish with higher infection intensity have less recipient contacts?

Subsetting the data frame down for this analysis

```
# subsetting down
IndContactsIndexinfrecip <- IndContactsIndexrecip %>%
  filter(InfectionTrt == "1")

# Scaling some of our variables for the analysis Index infection
# intensity
IndContactsIndexinfrecip$ScIndexWorm <- scale(IndContactsIndexinfrecip$IndexWorm)
```

2.3.1 Fitting our GLM for the recipient contacts by infection intensity

```
# Fitting the glmmTMB for index received contacts by infection
# intensity
IndexRCinflmrecip <- glmmTMB(RecipCRs ~ ScIndexWorm + I(ScIndexWorm^2) +
  recipsex + ScContactInitRs + I(ScContactInitRs^2) + ScIndexWorm:ScContactInitRs +
  ScIndexWorm:recipsex + I(ScIndexWorm^2):recipsex, family = beta_family(link = "logit"),
  data = IndContactsIndexinfrecip)

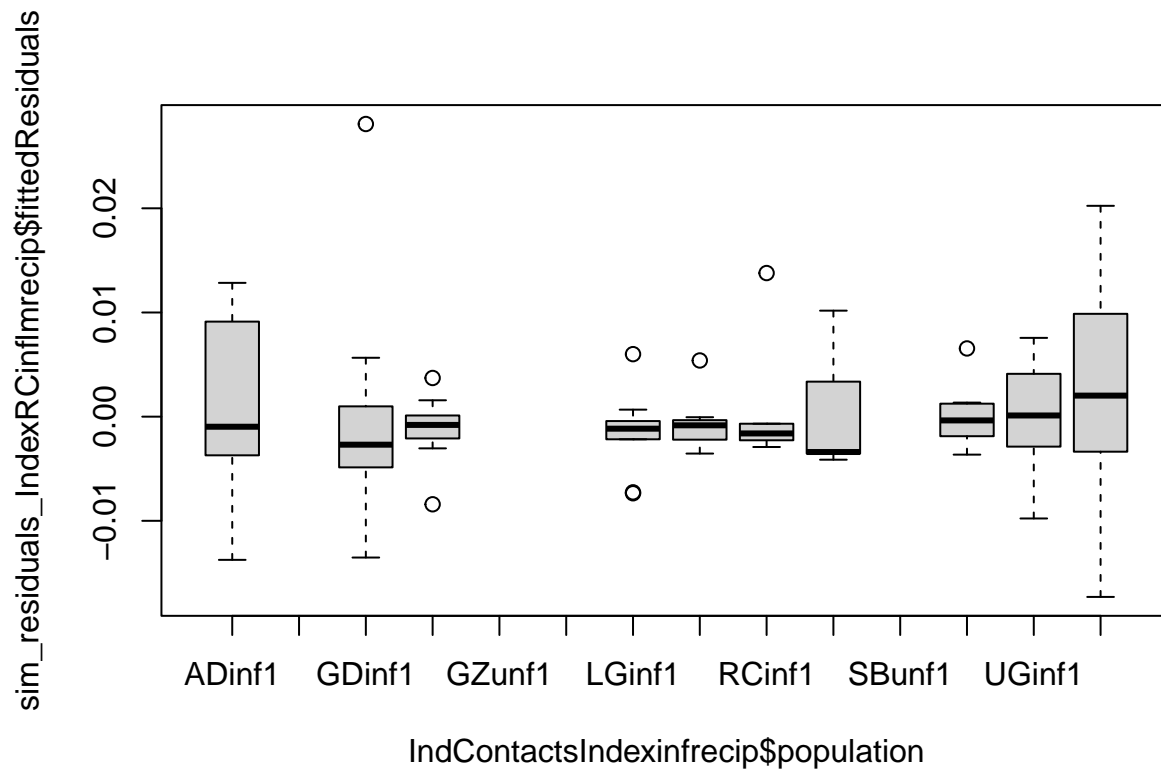
# summary of model
summary(IndexRCinflmrecip)
```

```
## Family: beta ( logit )
## Formula:
## RecipCRs ~ ScIndexWorm + I(ScIndexWorm^2) + recipsex + ScContactInitRs +
##      I(ScContactInitRs^2) + ScIndexWorm:ScContactInitRs + ScIndexWorm:recipsex +
##      I(ScIndexWorm^2):recipsex
## Data: IndContactsIndexinfrecip
##
##      AIC      BIC   logLik deviance df.resid
##   -714.2   -689.2    367.1   -734.2      80
##
##
## Dispersion parameter for beta family (): 506
##
## Conditional model:
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -4.78131    0.10348  -46.20 < 2e-16 ***
## ScIndexWorm     -0.12330    0.10519   -1.17  0.24112
## I(ScIndexWorm^2) -0.12916    0.05267   -2.45  0.01419 *
## recipsexM       0.50868    0.11469    4.44  9.2e-06 ***
## ScContactInitRs  1.37431    0.07492   18.34 < 2e-16 ***
## I(ScContactInitRs^2) 0.10111    0.04347    2.33  0.02004 *
## ScIndexWorm:ScContactInitRs 0.19357    0.09342    2.07  0.03825 *
## ScIndexWorm:recipsexM 0.36632    0.13452    2.72  0.00646 **
## I(ScIndexWorm^2):recipsexM -0.18727    0.09899   -1.89  0.05852 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

2.3.2 Validating the recipient contact rate by worm load model

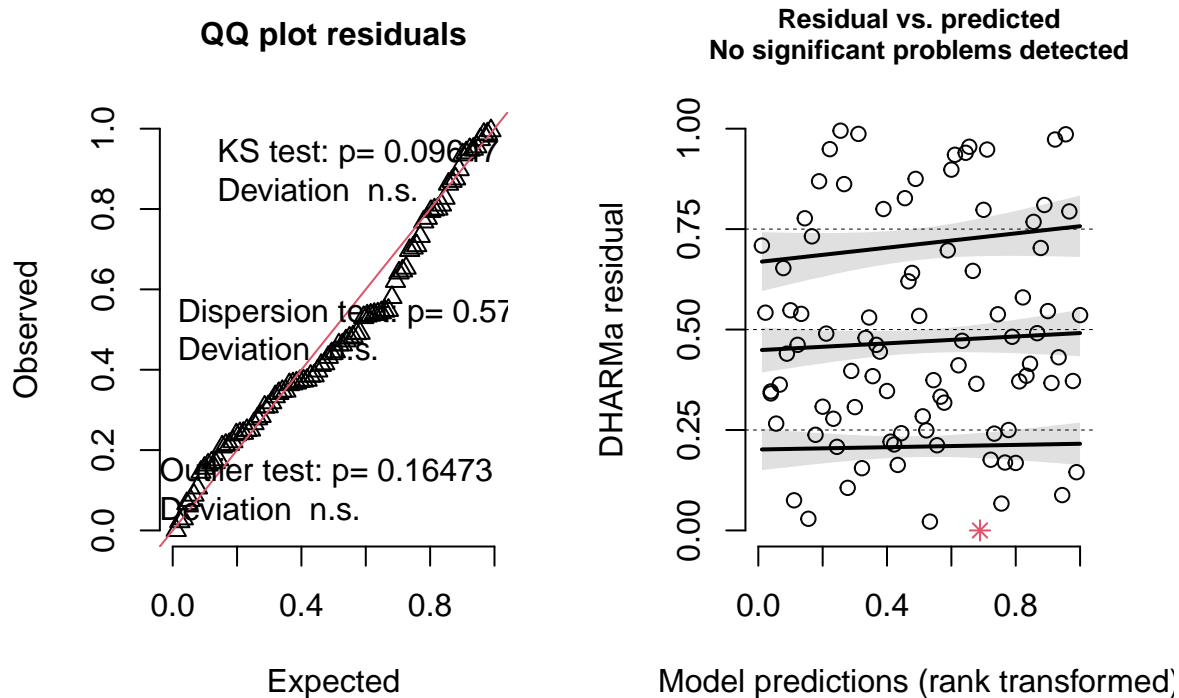
```
sim_residuals_IndexRCinflmrecip <- simulateResiduals(IndexRCinflmrecip,
  1000) #Quantile residuals

# Confirming there is no differences in median between groups
plot(sim_residuals_IndexRCinflmrecip$fittedResiduals ~ IndContactsIndexinfrecip$population)
```

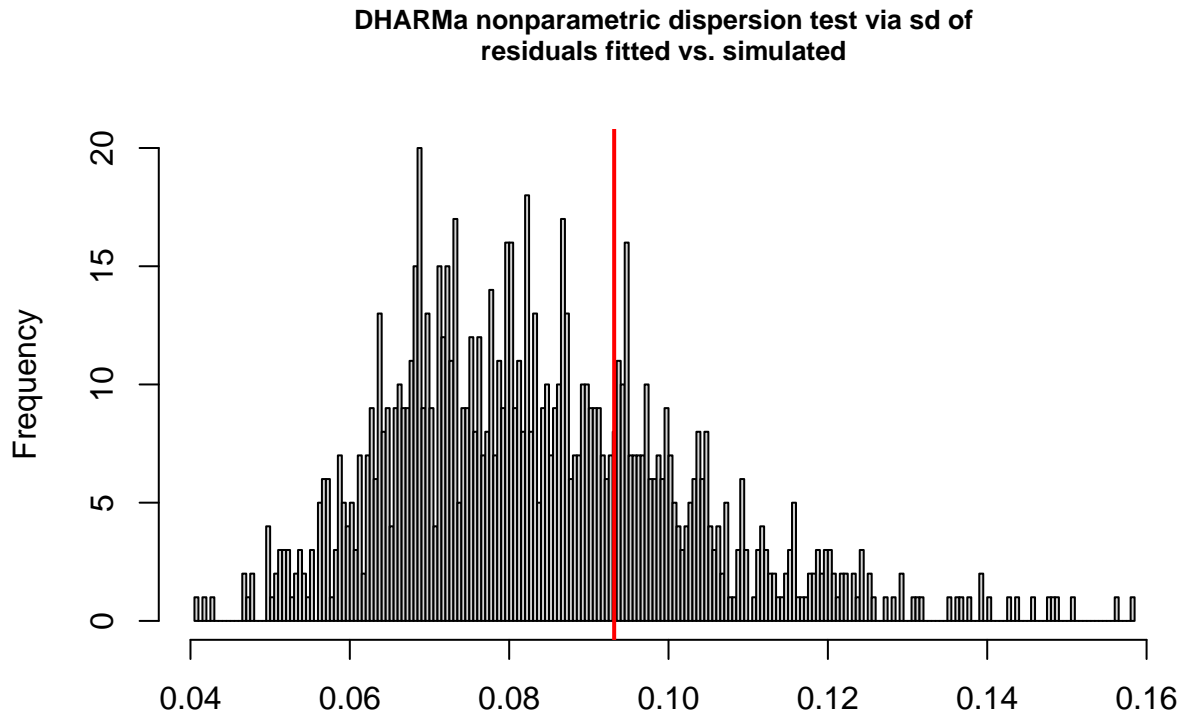


```
# Plotting the quantile residuals
plot(sim_residuals_IndexRCinflmrecip)
```

DHARMA residual



```
# Testing dispersion
testDispersion(sim_residuals_IndexRCinflmrecip)
```



Simulated values, red line = fitted model. p-value (two.sided) = 0.574

```
##
## DHARMA nonparametric dispersion test via sd of residuals fitted vs.
## simulated
##
## data: simulationOutput
## dispersion = 1.1119, p-value = 0.574
## alternative hypothesis: two.sided
```

```
# sim_residuals_IndexRCinflmininit
# <-simulateResiduals(IndexRCinflmininit, 1000) #Quantile residuals
# #Plotting the quantile residuals
# plot(sim_residuals_IndexRCinflmininit) #Testing dispersion
# testDispersion(sim_residuals_IndexRCinflmininit)
```

2.3.3 Using a likelihood ratio test to test for significance in recipient contact by index worm load model

```
drop1(IndexRCinflmrecip, test = "Chisq")
```

```
## Single term deletions
##
## Model:
```

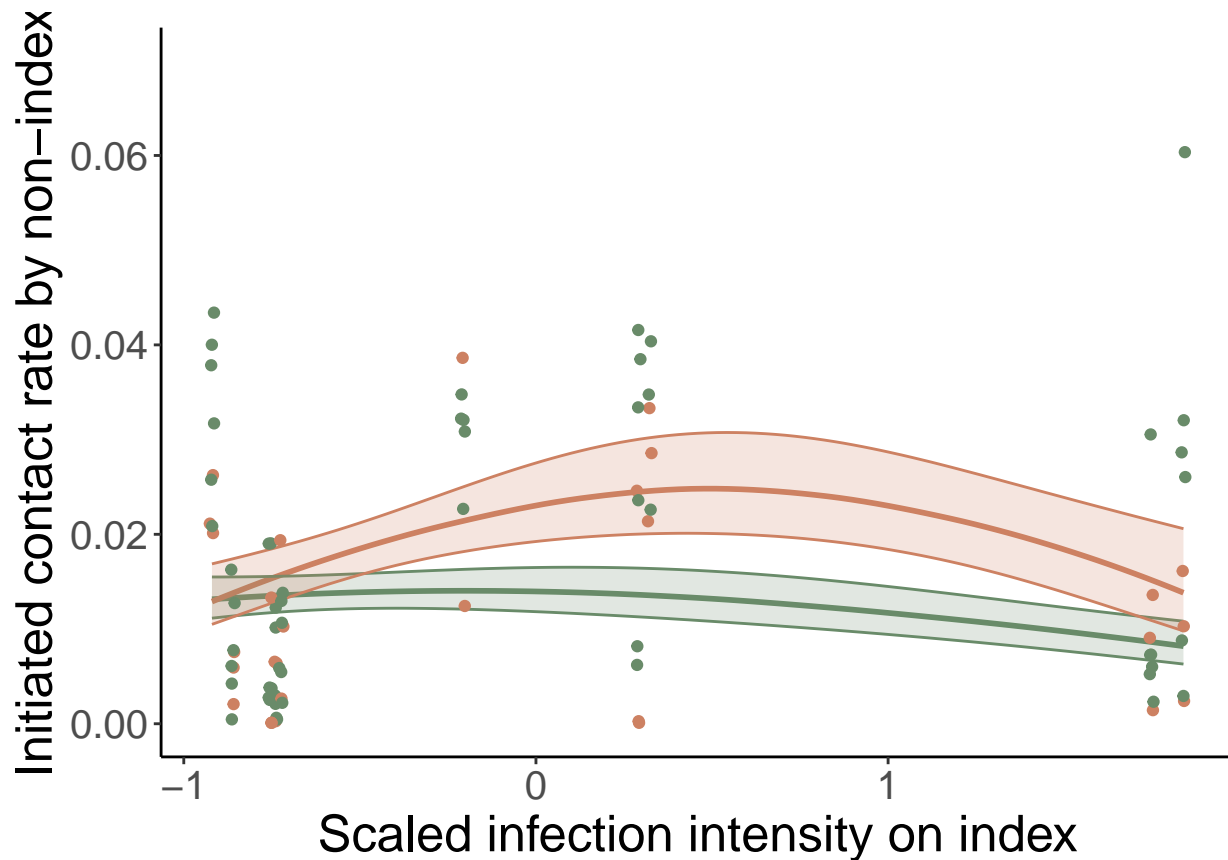
```
## RecipCRs ~ ScIndexWorm + I(ScIndexWorm^2) + recipsex + ScContactInitRs +
##      I(ScContactInitRs^2) + ScIndexWorm:ScContactInitRs + ScIndexWorm:recipsex +
##      I(ScIndexWorm^2):recipsex
##               Df      AIC      LRT Pr(>Chi)
## <none>                -714.22
## I(ScContactInitRs^2)      1 -711.54 4.6840 0.030446 *
## ScIndexWorm:ScContactInitRs 1 -711.70 4.5242 0.033419 *
## ScIndexWorm:recipsex      1 -708.94 7.2791 0.006976 **
## I(ScIndexWorm^2):recipsex  1 -712.71 3.5147 0.060827 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

2.3.4 Visualizing the relationship between recipient sex and recipient contact and worm load model

```
# Extracting fit and residuals from our model
IndexRCinflmVR <- visreg(IndexRCinflmrecip, "ScIndexWorm", "recipsex",
  scale = "response", partial = T, plot = FALSE)
# Extracting fit
IndexRCinflmVRfit <- IndexRCinflmVR$fit
# Extracting residuals
IndexRCinflmVRres <- IndexRCinflmVR$res
# Setting sex colorscheme
cpsex = c("darkseagreen4", "lightsalmon3")
# Plotting our relationship
ggplot(IndexRCinflmVRfit, aes(x = ScIndexWorm, y = visregFit, group = recipsex,
  fill = recipsex, color = recipsex)) + geom_smooth(method = "loess") +
  geom_ribbon(aes(ymin = visregLwr, ymax = visregUp), alpha = 0.2) +
  geom_jitter(data = IndContactsIndexinfrecip, aes(x = ScIndexWorm, y = RecipCRs)) +
  theme_classic() + xlab("Scaled infection intensity on index") + ylab("Initiated contact rate by non-")
  theme(text = element_text(size = 19)) + theme(legend.position = "none") +
  scale_fill_manual(values = cpsex) + scale_color_manual(values = cpsex) +
  ylim(0, 0.07)
```

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

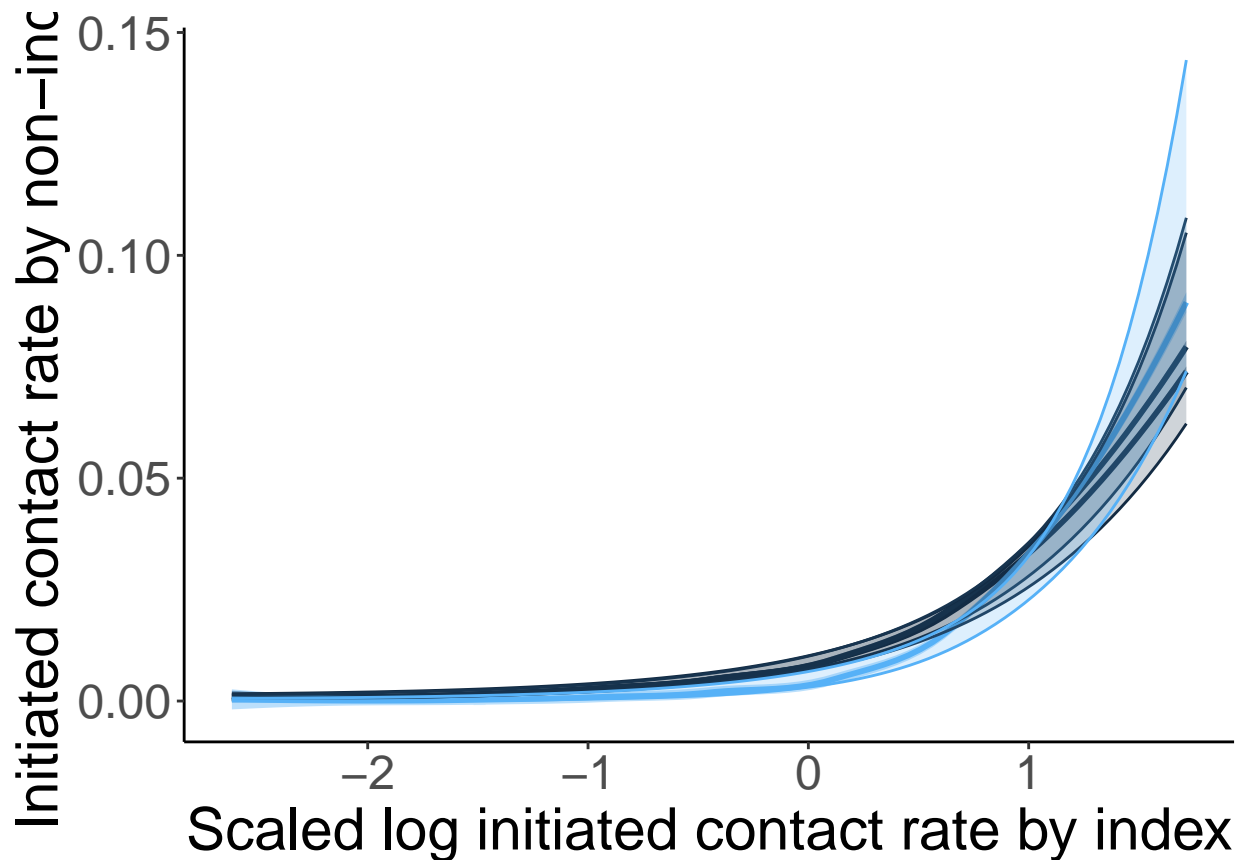
```
## Warning: Removed 5 rows containing missing values or values outside the scale range
## ('geom_point()').
```

```
# Extracting the fit and residuals from our model
InfIndexCR <- visreg(IndexRCinflmrecip, "ScContactInitRs", "ScIndexWorm",
  scale = "response", partial = T, overlay = TRUE, gg = TRUE, plot = FALSE)
# Extracting the fit of our model
InfIndexCRfit <- InfIndexCR$fit
# Extracting the residuals from the model
InfIndexCRres <- InfIndexCR$res

# Plotting the relationship between received contact rate, initiated
# contact rate, and infection intensity
ggplot(InfIndexCRfit, aes(x = ScContactInitRs, y = visregFit, group = ScIndexWorm,
  colour = ScIndexWorm, fill = ScIndexWorm)) + geom_smooth(data = InfIndexCRfit,
  aes(x = ScContactInitRs, y = visregFit)) + geom_ribbon(aes(ymin = visregLwr,
  ymax = visregUpr), alpha = 0.2) + theme_classic() + xlab("Scaled log initiated contact rate by index")
  ylab("Initiated contact rate by non-index") + theme(text = element_text(size = 22)) +
  theme(legend.position = "none")
```

```
## 'geom_smooth()' using method = 'loess' and formula = 'y ~ x'
```



2.3.5 Post hoc analysis for the popularity of index based on sex and index worms

```
# Subsetting down to females
IndContactsIndexinfrecipF <- IndContactsIndexinfrecip %>%
  filter(recipsex == "F")

# Subsetting down to males
IndContactsIndexinfrecipM <- IndContactsIndexinfrecip %>%
  filter(recipsex == "M")

# Fitting a GLMM for females only for post hoc analysis
IndexRCinflmrecipF <- glmmTMB(RecipCRs ~ ScIndexWorm + I(ScIndexWorm^2) +
  ScContactInitRs + I(ScContactInitRs^2) + ScIndexWorm:ScContactInitRs,
  family = beta_family(), data = IndContactsIndexinfrecipF)
# Summary for the female model model
summary(IndexRCinflmrecipF)
```

```
## Family: beta ( logit )
## Formula:
## RecipCRs ~ ScIndexWorm + I(ScIndexWorm^2) + ScContactInitRs +
## I(ScContactInitRs^2) + ScIndexWorm:ScContactInitRs
## Data: IndContactsIndexinfrecipF
##
##      AIC      BIC   logLik deviance df.resid
##   -453.4   -438.8    233.7   -467.4       53
```

```
##
##
## Dispersion parameter for beta family (): 476
##
## Conditional model:
##
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)      -4.76123    0.11525  -41.31  <2e-16 ***
## ScIndexWorm       -0.16894    0.13282   -1.27   0.2034
## I(ScIndexWorm^2)  -0.13398    0.05521   -2.43   0.0152 *
## ScContactInitRs    1.34848    0.13756    9.80  <2e-16 ***
## I(ScContactInitRs^2) 0.10363    0.08248    1.26   0.2089
## ScIndexWorm:ScContactInitRs 0.25914    0.12749    2.03   0.0421 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# Likelihood ratio test for significance in the female model
drop1(IndexRCinflmrecipF, test = "Chisq")

## Single term deletions
##
## Model:
## RecipCRs ~ ScIndexWorm + I(ScIndexWorm^2) + ScContactInitRs +
##           I(ScContactInitRs^2) + ScIndexWorm:ScContactInitRs
##
##               Df      AIC      LRT Pr(>Chi)
## <none>                -453.41
## I(ScIndexWorm^2)       1 -449.79 5.6269 0.01769 *
## I(ScContactInitRs^2)   1 -453.99 1.4190 0.23356
## ScIndexWorm:ScContactInitRs 1 -450.91 4.5031 0.03383 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# Fitting a GLMM for male post hoc analysis
IndexRCinflmrecipM <- glmmTMB(RecipCRs ~ ScIndexWorm + I(ScIndexWorm^2) +
  ScContactInitRs + I(ScContactInitRs^2) + ScIndexWorm:ScContactInitRs,
  , family = beta_family(), data = IndContactsIndexinfrecipM)
# Summary for the male model
summary(IndexRCinflmrecipM)

## Family: beta ( logit )
## Formula:
## RecipCRs ~ ScIndexWorm + I(ScIndexWorm^2) + ScContactInitRs +
##           I(ScContactInitRs^2) + ScIndexWorm:ScContactInitRs
## Data: IndContactsIndexinfrecipM
##
##      AIC      BIC    logLik deviance df.resid
##  -256.9   -247.1    135.5   -270.9      23
##
##
## Dispersion parameter for beta family (): 668
##
## Conditional model:
##
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)      -4.390565    0.139093 -31.566  < 2e-16 ***
```

```
## ScIndexWorm          0.280794    0.105600    2.659 0.007837 **
## I(ScIndexWorm^2)     -0.295458    0.086609   -3.411 0.000646 ***
## ScContactInitRs      1.494855    0.124637   11.994 < 2e-16 ***
## I(ScContactInitRs^2) 0.150132    0.079416    1.890 0.058698 .
## ScIndexWorm:ScContactInitRs -0.009143    0.156285   -0.059 0.953349
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# Likelihood ratio test for significance in the male
drop1(IndexRCinflmrecipM, test = "Chisq")
```

```
## Single term deletions
##
## Model:
## RecipCRs ~ ScIndexWorm + I(ScIndexWorm^2) + ScContactInitRs +
##       I(ScContactInitRs^2) + ScIndexWorm:ScContactInitRs
##              Df      AIC      LRT Pr(>Chi)
## <none>                -256.95
## I(ScIndexWorm^2)       1 -249.41  9.5334 0.002018 **
## I(ScContactInitRs^2)   1 -255.60  3.3430 0.067491 .
## ScIndexWorm:ScContactInitRs 1 -258.94  0.0034 0.953382
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

2.4 Do groups with infected versus uninfected index fish change their contact rate different across days 2 and 3 and does this vary by sex?

Subsetting down to the data we need for this analysis

```
IndContacts4ch <- IndContacts4 %>%
  distinct(fishID2, recipID2, .keep_all = TRUE)
# Subsetting down to females
IndContacts4chF <- IndContacts4ch %>%
  filter(Sex == "F")
# Subsetting down to males
IndContacts4chM <- IndContacts4ch %>%
  filter(Sex == "M")
```

We are fitting a linear mixed model with change in contact rate between dyas 2 and 3 for all uninfected fish in the experiment.

2.4.1 Fitting our linear model for the change in behavior across days 2 and 3

```
# fitting glmmTMB for total contacts over sex and infection treatment
Contactsmodel <- lmer(ChContactinit ~ InfectionTrt + Sex + recipsex + (1 |
  population) + (1 | fishID2), data = IndContacts4ch)

# Summary for our model
summary(Contactsmodel)
```

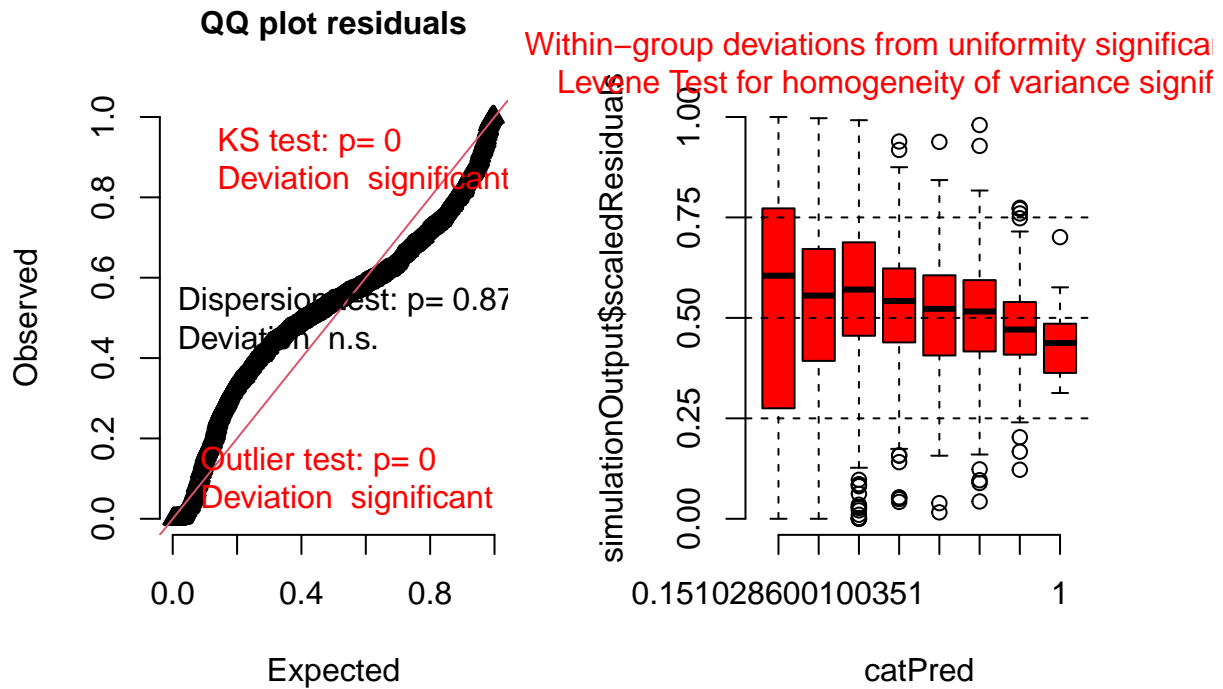
```

## Linear mixed model fit by REML ['lmerMod']
## Formula: ChContactinit ~ InfectionTrt + Sex + recipsex + (1 | population) +
##      (1 | fishID2)
##      Data: IndContacts4ch
##
## REML criterion at convergence: -5040.4
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -4.8411 -0.3423  0.0428  0.4171  6.0639
##
## Random effects:
##      Groups      Name      Variance Std.Dev.
## fishID2      (Intercept) 5.565e-05 0.007460
## population (Intercept) 9.093e-05 0.009536
## Residual              3.314e-04 0.018204
## Number of obs: 1008, groups: fishID2, 126; population, 14
##
## Fixed effects:
##              Estimate Std. Error t value
## (Intercept)  -0.002680   0.005101  -0.525
## InfectionTrt1 -0.006581   0.005967  -1.103
## SexM          0.003802   0.001868   2.035
## recipsexM     0.005012   0.001226   4.088
##
## Correlation of Fixed Effects:
##              (Intr) Infct1 SexM
## InfectnTrt1 -0.835
## SexM         -0.129  0.000
## recipsexM    -0.090  0.000  0.082

sim_residuals_Contactsmodel <- simulateResiduals(Contactsmodel, 1000) #Quantile residuals
# Plotting the quantile residuals
plot(sim_residuals_Contactsmodel)

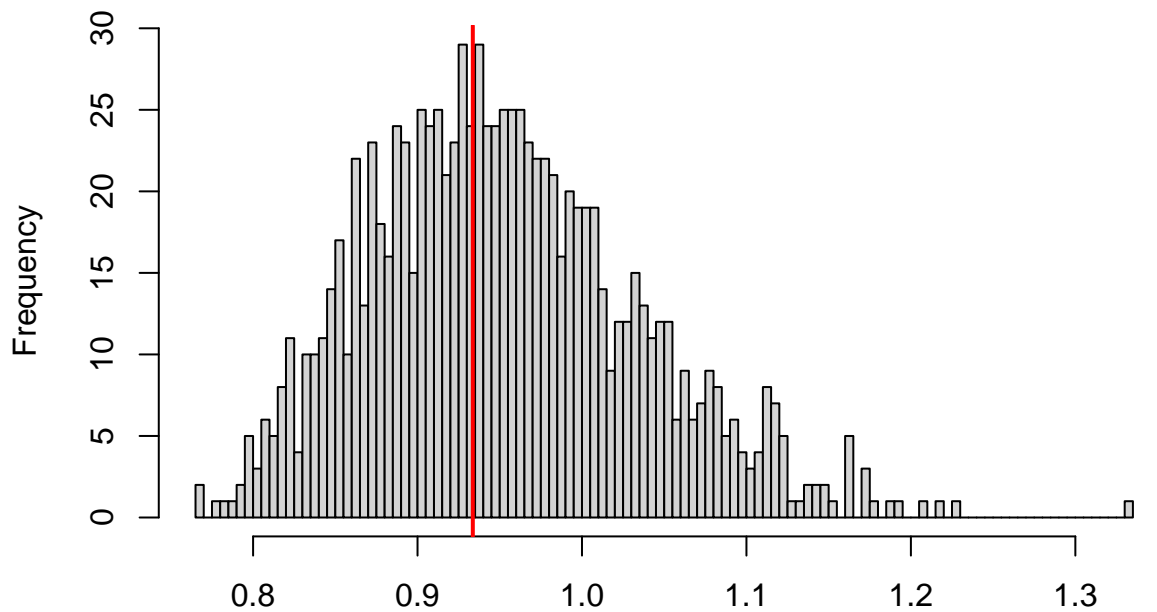
```

DHARMA residual



```
# Testing dispersion
testDispersion(sim_residuals_Contactsmodel)
```

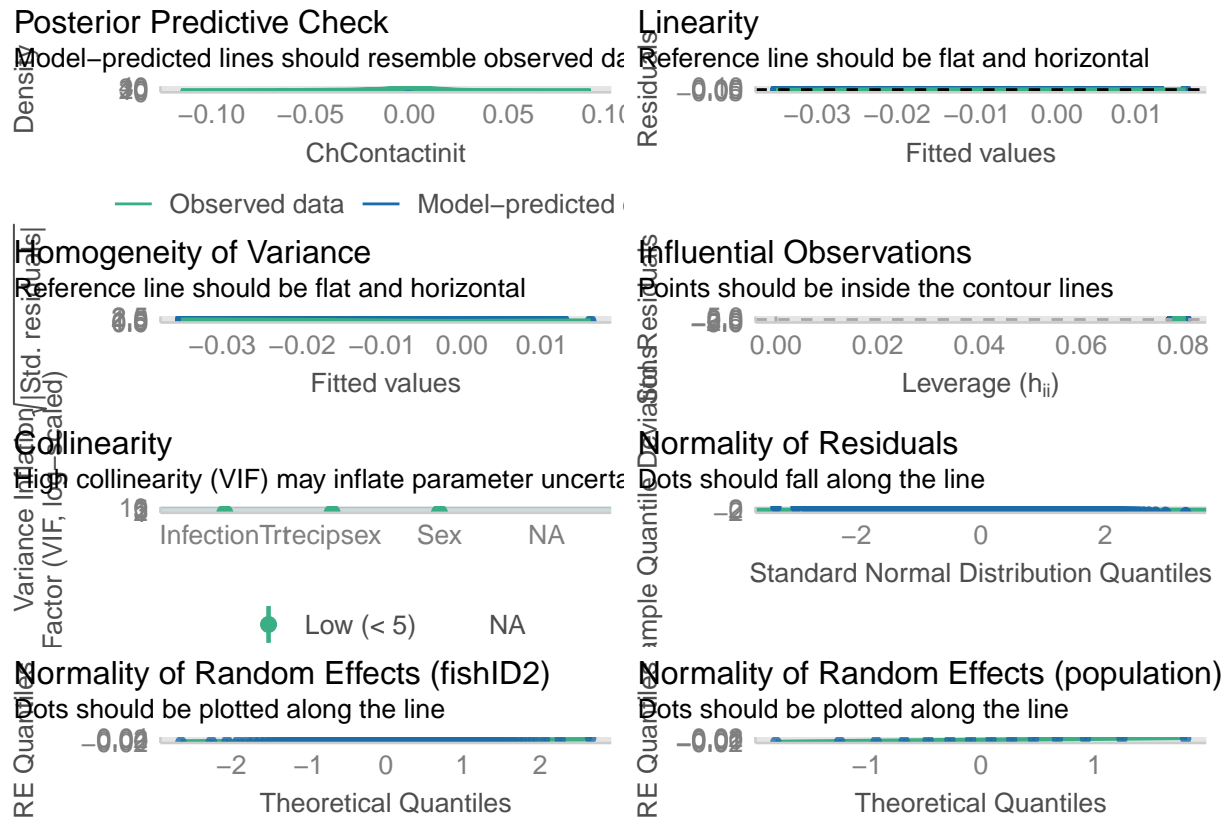
DHARMA nonparametric dispersion test via sd of residuals fitted vs. simulated



Simulated values, red line = fitted model. p -value (two.sided) = 0.878

```
##
## DHARMA nonparametric dispersion test via sd of residuals fitted vs.
## simulated
##
## data: simulationOutput
## dispersion = 0.98044, p-value = 0.878
## alternative hypothesis: two.sided
```

```
check_model(Contactsmodel)
```



```
Anova(Contactsmodel, type = "2", test = "F")
```

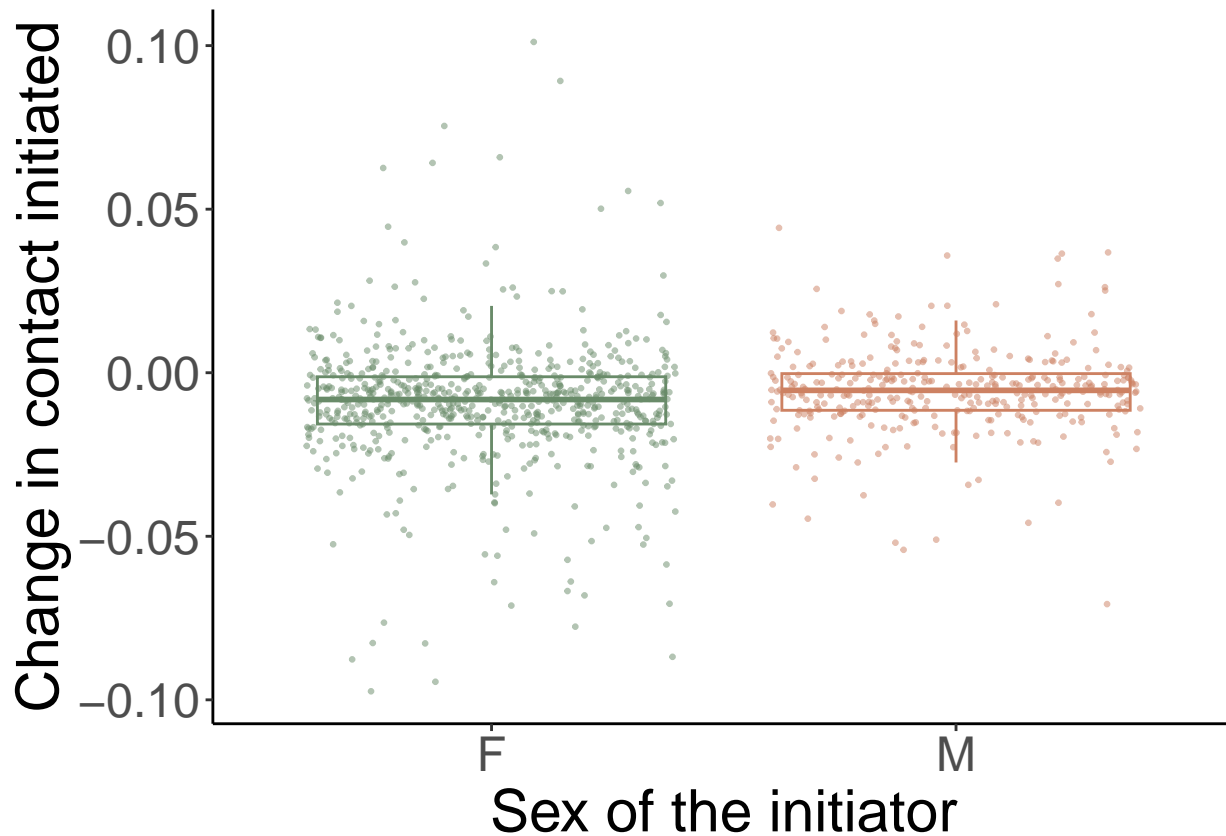
```
## Analysis of Deviance Table (Type II Wald F tests with Kenward-Roger df)
##
## Response: ChContactinit
##              F Df Df.res    Pr(>F)
## InfectionTrt  1.2165  1  12.00  0.2917
## Sex           4.1414  1 112.51  0.0442 *
## recipsex     16.7139  1 881.00 4.743e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# Anova(ContactsmodelF, type='2', test='F') Anova(ContactsmodelM,
# type='2', test='F')
```

```

# Setting a labeller to change axis names in the plot
labellerI <- labeller(InfectionTrt = c(`1` = "Infected Treatment", `0` = "Uninfected Treatment"))
# Setting colorscheme for infection and sex categorical variables
cpinf = c("darkgray", "purple")
cpsex = c("darkseagreen4", "lightsalmon3")
# Exporting the fit and residual for sex in our model
ContactsmodelVR <- visreg(Contactsmodel, "Sex", scale = "response", partial = T,
  plot = FALSE)
# Extracting fit
ContactsmodelVRfit <- ContactsmodelVR$fit
# Extracting residuals
ContactsmodelVRres <- ContactsmodelVR$res
# Plotting out the relationship
ggplot(ContactsmodelVRres, aes(x = Sex, y = visregRes, color = Sex)) +
  geom_boxplot(aes(color = Sex), outliers = FALSE) + geom_jitter(alpha = 0.5,
    size = 0.5) + theme_classic() + theme(text = element_text(size = 22)) +
  ylab("Change in contact initiated") + xlab("Sex of the initiator") +
  scale_color_manual(values = cpsex) + theme(legend.position = "none")

```



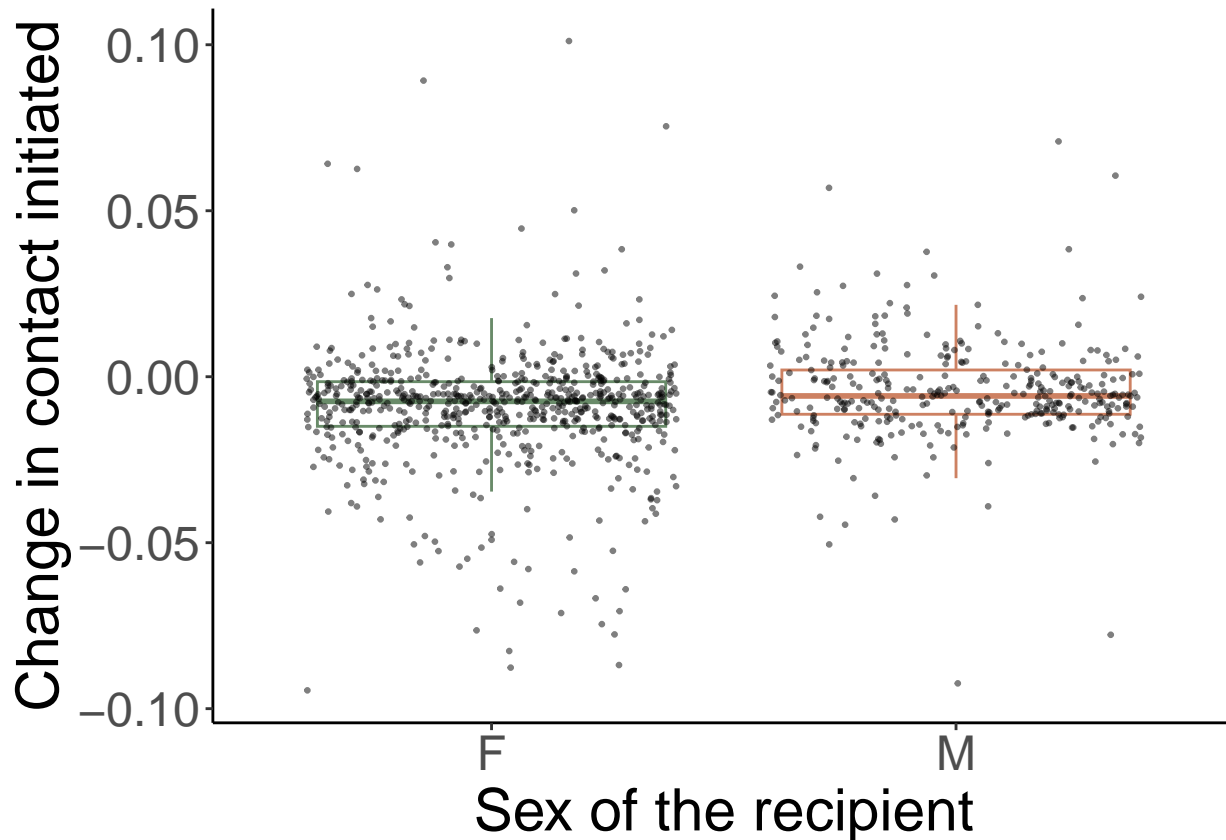
```

# Extracting residuals and fit for the recipient sex in the model
ContactsmodelrecipVR <- visreg(Contactsmodel, "recipsex", scale = "response",
  partial = T, plot = FALSE)
# Extracting fit
ContactsmodelrecipVRfit <- ContactsmodelrecipVR$fit
# Extracting residuals
ContactsmodelrecipVRres <- ContactsmodelrecipVR$res

```



```
# Plotting the residuals
ggplot(ContactsmodeIrecipVRres, aes(x = recipsex, y = visregRes)) + geom_boxplot(aes(color = recipsex),
  outliers = FALSE) + geom_jitter(alpha = 0.5, size = 0.5) + theme_classic() +
  theme(text = element_text(size = 22)) + ylab("Change in contact initiated") +
  xlab("Sex of the recipient") + scale_color_manual(values = cpsex) +
  theme(legend.position = "none")
```



2.4.2 Post-hoc analyses for exploring sex differences in the change in contact rate between days 2 and 3

```
# Fitting our post-hoc analysis for female change in contact model
ContactsmodeIF <- lmer(ChContactinit ~ (1 | population) + (1 | fishID2),
  data = IndContacts4chF)
```

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

```
# Summary for the female model
summary(ContactsmodeIF)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: ChContactinit ~ (1 | population) + (1 | fishID2)
## Data: IndContacts4chF
##
## REML criterion at convergence: -3278.5
```

```
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -4.5939 -0.3498  0.0270  0.4104  5.5912
##
## Random effects:
##   Groups      Name      Variance Std.Dev.
## fishID2      (Intercept) 0.0000000 0.00000
## population (Intercept) 0.0001584 0.01259
## Residual                0.0004134 0.02033
## Number of obs: 672, groups: fishID2, 84; population, 14
##
## Fixed effects:
##              Estimate Std. Error t value
## (Intercept) -0.005501   0.003454  -1.593
## optimizer (nloptwrap) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')
```

```
# One sample t-test to test if the change in contact rate is
# different than zero
t.test(IndContacts4chF$ChContactinit)
```

```
##
## One Sample t-test
##
## data: IndContacts4chF$ChContactinit
## t = -6.0216, df = 671, p-value = 2.845e-09
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## -0.007294344 -0.003707031
## sample estimates:
## mean of x
## -0.005500688
```

```
# Fitting our post-hoc analysis for male change in contact model
ContactsmodelM <- lmer(ChContactinit ~ (1 | population) + (1 | fishID2),
  data = IndContacts4chM)
```

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

```
# Summary for the male model
summary(ContactsmodelM)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: ChContactinit ~ (1 | population) + (1 | fishID2)
## Data: IndContacts4chM
##
## REML criterion at convergence: -1859.1
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -4.4324 -0.3528 -0.0027  0.3729  4.0355
```

```
##
## Random effects:
##   Groups      Name      Variance Std.Dev.
## fishID2      (Intercept) 0.0001385 0.01177
## population (Intercept) 0.0000000 0.00000
## Residual                0.0001754 0.01325
## Number of obs: 336, groups: fishID2, 42; population, 14
##
## Fixed effects:
##               Estimate Std. Error t value
## (Intercept) -0.002325   0.001954   -1.19
## optimizer (nloptwrap) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')

# One sample t-test to test if the change in contact rate is
# different than zero
t.test(IndContacts4chM$ChContactinit)

##
## One Sample t-test
##
## data: IndContacts4chM$ChContactinit
## t = -2.4168, df = 335, p-value = 0.01619
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
##  -0.0042177269 -0.0004326831
## sample estimates:
##   mean of x
## -0.002325205
```