Third Fitting of Thermal Models

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Goal

Fit series of thermal models, including, ultimately, those found in rTPC to data collected in the Derryberry lab.

Recap

First and Second Fitting

- Temp is hard to control in chambers, so we should work with temp_mean (mean value during trial), not temp_target.
- humidity_mean and temp_mean strongly co-vary so consider using aggregate index as explanatory variable
 - Vapor Pressure Deficit (provided by humidity package)
 - Heat Index (formulated for humans, and provided by weathermetrics)
- Round 3 data only for curve fitting
 - Only round where temp_mean and humidity_mean exist.
- count_total_round are consistent between Rounds 1 and 3, so could use that info to classify birds, estimate variance function in response, etc.
- One bird in round = 3 collapsed during the trial so it was terminated. We've set the song_count from NA to 0 and should consider making at temp_working column using temp_target in this instance and temp_mean in all other instances. Would need to do something similar for relative humidity, i.e. use mean(humidity_mean) for the temp_target value.

Third Fitting

- Created temp and humidity variables from temp_mean and humidity_mean for males who didn't collapse, used mean(temp_mean) and mean(humidity_mean) values for one male that did.
- Copied data_full and set male = "combined" so we could look at all of the data at once.
- Learned about using model argument for glm models
- Can't fit ME models
 - I don't understand why the ME model with song_count where we use a RE for the intercept doesn't have a similar effect as using song_prop. This intercept value is essentially multiplying by a constant, so we could try and specify this value using offset(log(count_total_round)). So why doesn't this work?

Fourth Fitting

Preliminary Notes

Which Explanatory Variable: temp, humidity, vpd?

- Liz will look into operative temp
- Use of vpd_mean is somewhat arbitrary since it depends on the experimental design.
 - Use of 0 reference in vpd seems less arbitrary than using 0 C.
- Note that we could scale temp_mean relative to C = 45, which is the understood thermal maximum.
 - This is what Liz wants to do since it's easier to interpret

Type of Model Fitted

- Use temp_mean temp_ref where Don't worry about random effects for now
- Try the quadratic fit with just temperature and the 0 set at 45C.
 - Quadratic function pull out the peak, the curvature at the peak, calculate the intercept at some temp – what is the intercept
- Begin fitting rTPC models to combined dataset.

Summary of Results

- Tried fitting lots of variations of glm and glmer
- Optimizer matters in terms of convergence.
- Adding fixed male term rather than treating it as RE via (1|male) was very important since the variation in overall male singing varies so greatly.
- poisson fits indicate data is greatly overdispersed.
- Got one particular form of GLMM (glmer) working well (i.e. when there's no covariance matrix). I think this is a valid approach, but am unsure and a bit concerned since the | fitting shows the RE for x1 and $x1^2$ being close to perfectly correlated.
- Overall results indicate taht this approach is problematic
 - Data is extremely overdispersed in these fits.
 - As a result, no signal beyond a male effect on overall activity is reliably detected.

Future Steps

Revisit lmList

- are the estiamtes from individual models consistent with the paramters estimated with the | and /or | | models? This is really only worth doing if
 - We think we want to ultimately use the quadratic link = "log" models
 - We want to make sure we correctly understand GLMM

Analyze with rTPC

According to Padfield et al. (2021)

However, [the rTPC] pipeline does not accommodate non-independent (related) replicates, and clustered or stratified sampling (possibly with missing values). In such situations, nonlinear mixed effects model fitting (e.g. using the nlme r package; Oddi et al., 2019) or Bayesian approaches (e.g. using the brms r package; Bürkner, 2017) would be more appropriate. Nevertheless, for fitting massive TPC datasets to multiple mathematical models, rTPC offers a simple, reliable and reproducible computational pipeline with robust methods for calculation of model uncertainty, requiring minimal statistical and computational expertise, and suitable for a wide range of applications.

Thus, we can't incorporate individual or random effects. We could try and use all of the observations in song_prop to deal with the male term.

Analyze with other packages

• i.e. use stan and/or brms which uses stan.

Including beak and mass data

- Include beak size (surface area) as additional explanatory variable. We do have bird mass as well
- Notes from Liz
 - Do birds with larger beaks maintain singing at higher rates at higher temps?
 - We calculated bill surface area approximately as the surface area of a cone: length * pi * (width + depth) / 4.
 - Because larger animals produce more heat (Kleiber 1932), we scaled bill size relative to heat production by dividing bill surface area by expected daily energy consumption (mass0.658; Speakman and Kr´ol 2010, Hudson et al. 2013).
 - So, looks like we need to calculate bill surface area and then scale relative to heat production (bill surface area/mass0.658)

Using Data from Rounds 1 and 2

- Liz: Ask Kayci about temp_mean and humidity_mean data for rounds 1 and 2.
- Look for order effects in round 1 and 2? Can we use any of this data? Seems like there will be an issue if we throw out, say, first two trials, when using song_prop. Note that since total_count is consistent for a bird between rounds, this may not be an issue.
- Begin fitting rTPC models to combined dataset.
- We could include round = 2
 - Would need to down weight song_count values when combining across count_total_round values.

Additional Liz

- Follow up with Ray/Juan about ground versus surface temp for operative temperature.
- What is the 0 for operative temperature?
- Ponder utility of other ZF data on panting.

Set up

Install libraries

```
## install packages user might not have by replacing FALSE with TRUE
if(FALSE) {
    BiocManager::install("mixOmics") ## needed by RVAideMemoire
    install.packages(c("RSQLite", "nls.multstart", "lme4", "RVAideMemoire"))
    ## Install the thermal curve package from git_hub, not cran
    remotes::install_github("padpadpadpad/rTPC")
}
## load libraries
library(stats)
require(MASS) # provides negative binomial fitting: glm.nb
## Loading required package: MASS
library(RSQLite) # Don't think we need this.
library(rTPC) ##
library(nls.multstart)
library(broom)
library(tidyverse)
## -- Attaching packages
## -----
## tidyverse 1.3.2 --
## v ggplot2 3.3.6 v purrr 0.3.4

## v tibble 3.1.8 v dplyr 1.0.99.9000

## v tidyr 1.2.0.9000 v stringr 1.4.1

## v readr 2.1.3 v forcats 0.5.2
## v readr 2.1.3
                             v forcats 0.5.2
## -- Conflicts -----
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
## x dplyr::select() masks MASS::select()
require(ggplot2)
require(ggpubr)
## Loading required package: ggpubr
require(viridisLite)
## Loading required package: viridisLite
```

```
#options(ggplot2.continuous.colour="viridis",
         qqplot2.discrete.colour="viridis",
#
         ggplot2.scale_fill_discrete = scale_fill_viridis_d,
#
         ggplot2.scale_fill_continuous = scale_fill_viridis_c)
require(GGally)
## Loading required package: GGally
## Registered S3 method overwritten by 'GGally':
    method from
##
     +.gg
            ggplot2
require(reshape2)
## Loading required package: reshape2
## Attaching package: 'reshape2'
## The following object is masked from 'package:tidyr':
##
##
       smiths
require(lme4)
## Loading required package: lme4
## Loading required package: Matrix
##
## Attaching package: 'Matrix'
##
## The following objects are masked from 'package:tidyr':
##
##
       expand, pack, unpack
require(RVAideMemoire) # provides overdisp.glmer()
## Loading required package: RVAideMemoire
## *** Package RVAideMemoire v 0.9-81-2 ***
##
## Attaching package: 'RVAideMemoire'
## The following object is masked from 'package:lme4':
##
##
       dummy
## The following object is masked from 'package:broom':
##
##
       bootstrap
require(humidity) ## provides VPD
## Loading required package: humidity
```

```
require(weathermetrics)

## Loading required package: weathermetrics

require(latex2exp)

## Loading required package: latex2exp
```

Load Data

```
## Read in ZEFI Data sets
## Treat 'repeatability' as round = 0
## Add round info
## Repeatability was done between round 1 and 2, female was present, but only one temp. so treating as
git_root <- system("git rev-parse --show-toplevel", intern = TRUE)</pre>
data_raw = list()
data_raw[[1]] <- read.csv(file.path(git_root, "data", "raw_data", "HSPi-Round-1-Heat-Trials.csv")) %>% 1
    ## Note T237 and T230 are missing numbers in the song_count column
    ## so we are filtering these observations out until they are found
   filter(!is.na(song_count))
data_raw[[2]] <- read.csv(file.path(git_root, "data", "raw_data", "HSPi-Repeatability-Song-Count.csv"))
   mutate(round = 2) %>%
   group_by(male) %>%
   mutate(test_order = rank(date)) %>%
   ungroup()
data_raw[[3]] <-read.csv(file.path(git_root, "data", "raw_data", "HSPi-Round-2-Heat-Trials.csv")) %%
   mutate(round = 3) %>%
    ## Deal with missing temp_mean and humidity_mean values
   ## in round == 3
    ## 2022/10/19 - code no longer needed
    ## group_by(temp_target) %>%
    ##mutate(temp = if_else((round == 3 & is.na(temp_mean)),
                            mean(temp_mean, na.rm = TRUE),
                            temp mean)) %>%
    ##mutate(humidity = if_else((round == 3 & is.na(humidity_mean)),
    ##
                                mean(humidity_mean, na.rm = TRUE),
    ##
                                humidity_mean)) %>%
    ungroup()
## Join data and discard empty columns
data_full <- full_join(data_raw[[1]], data_raw[[2]]) %>%
   full_join(data_raw[[3]]) %>%
   discard(~all(is.na(.) | . =="")) %>% ## get rid of columns of only NA
```

```
mutate(trial_completed = !(is.na(song_count)) ) %>%
mutate(song_count = ifelse(is.na(song_count), 0, song_count)) %>%
mutate(song_count = song_count*1.0) %>% ## convert to a double so it's not treated as an integer
mutate(chamber = as.factor(chamber), male = as.factor(male)) %>%
## create a global variable trial_order based on individual rounds
mutate(trial_index = as.integer(round*10+test_order)) %>%
mutate(song_count_plus_1 = (song_count + 1)) %>%
mutate(log_song_count_plus_1 = log(song_count + 1)) %>%
mutate(temp_target = as.numeric(temp_target)) %>%
## Add column with total song_count for a given round
group_by(male, round ) %>%
mutate(count_total_round = sum(song_count) ) %>%
ungroup() %>%
mutate(song_prop = song_count/count_total_round) %>%
## assuming poisson error
## From glm man page
## > Non-'NULL' 'weights' can be used to indicate that different
## > observations have different dispersions (with the values in
## > 'weights' being inversely proportional to the dispersions);
## add +1 to deal with single 0
mutate(count_wt = 1/(song_count + 1)) %>%
## need to rescale wts for song_prop data
mutate(prop_wt = count_wt * count_total_round^2) %>%
## Add vpd
mutate(svp = SVP(t = temp_mean + 273.15, isK = TRUE), vpd = svp*(1-humidity_mean/100)) %>%
group_by(round) %>%
mutate(vpd_offset = vpd - mean(vpd)) %>%
ungroup() %>%
relocate(song_count, song_prop, vpd, temp_mean, humidity_mean, .after = male) %>%
mutate() ## Dummy function so we can comment out lines above it w/o any issues
```

Joining, by = c("male", "chamber", "date", "song_count", "counter", "test_order", "temp_target", "ro ## Joining, by = c("male", "chamber", "date", "song_count", "counter", "test_order", "temp_target", "ro

Fourth Analysis

Examine Data

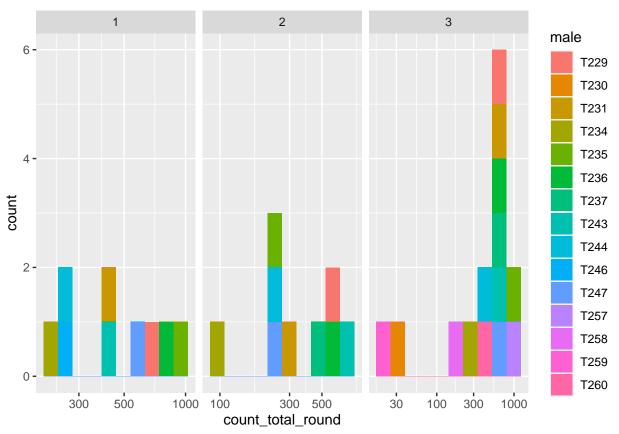
```
data_count_total <- data_full %>% group_by(round) %>%
    select(male, round, count_total_round) %>%
    distinct()

t <- ggplot(data_count_total, aes(count_total_round, fill = male)) +
    geom_histogram(bins = 10) +
    scale_x_log10()

hist_count_total <- t +
    facet_grid(cols =vars(round), scales = "free_x")
hist_count_total</pre>
```

Warning: Transformation introduced infinite values in continuous x-axis

Warning: Removed 1 rows containing non-finite values (stat_bin).



Compare count_total_round between round 1 and 3

• See third.fitting.Rmd

Result

- As before, we see strong consistancy between round 1 and 3.
- Consistency with round 2 is weaker, but sample sizes are smaller: 3 trials/male in round 2 vs 6 trials/male in round 3.

Create & Plot Filtered Data

```
data_ind <- data_full %>%
    filter(round==3) %>%
    filter(count_total_round >=150)

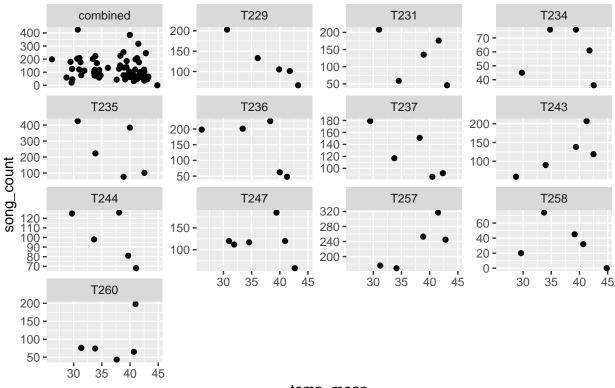
## copy data frame and assign `male = "combined")
data_comb <- data_ind %>% mutate(male = "combined")

data <- bind_rows(data_ind, data_comb)

xlab <- "Temperature"
ylab <- "song_count"</pre>
```

```
plot_temp_data <-
    ggplot(data) +
    aes(x = temp_mean,
        y = song_count) +
    facet_wrap("male", scales = "free_y") +
    geom_point() +
labs( title = paste( ylab, " vs ", xlab))
last_plot()</pre>
```

song_count vs Temperature



temp_mean

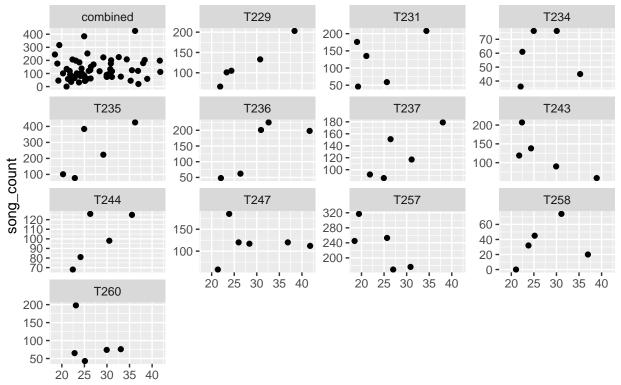
```
## Create again for humidity

xlab <- "humidity"

plot_humidity_data <-
    ggplot(data) +
    aes(x = humidity_mean,
        y = song_count) +
    facet_wrap("male", scales = "free_y") +
    geom_point() +

labs( title = paste( ylab, " vs ", xlab))
last_plot()</pre>
```

song_count vs humidity



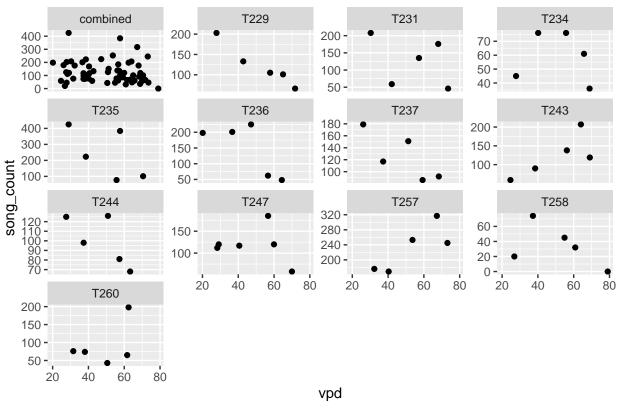
humidity_mean

```
## Create for vpd

xlab <- "vpd"

plot_vpd_data <-
    ggplot(data) +
   aes(x = vpd,
        y = song_count) +
   facet_wrap("male", scales = "free_y") +
   geom_point() +
   labs( title = paste( ylab, " vs ", xlab))
last_plot()</pre>
```

song_count vs vpd

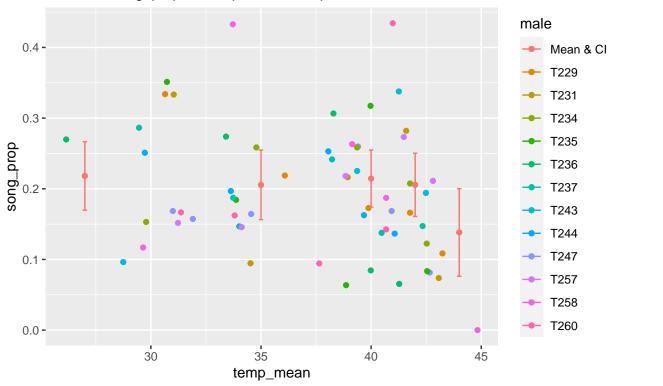


Work with mean and song_prop

• This approach won't work because temp_target varies greatly from temp_mean.

```
plot_combine <- ggplot(data_ind) +</pre>
                  aes(x = temp_mean, y = song_prop, color = male) +
                  geom_point()
# last_plot()
data_summarize <-
    data_ind %>% group_by(temp_target) %>%
    summarize(song_mean = mean(song_prop),
              song_sd = sd(song_prop),
              song_n= dplyr::n(),
              song_se = song_sd/sqrt(song_n),
              song_ci = song_se*1.96
              )
plot_summarize <- ggplot(data_summarize) +</pre>
    aes(x = temp_target, y = song_mean) +
    geom_point()+
    geom_errorbar(aes(ymin=song_mean - song_ci, ymax=song_mean+song_ci), width=.2)
# last_plot()
```

Mean(song_prop) vs. temp_target with 95% CI for mean Individual song_prop vs. temp_mean also plotted



Result

- Data is noisy.
- $\bullet~$ Using ${\tt song_prop}$ reduces impact of male T225 on lower temps.
- Note temp_target is not necessarily accurate.

Humidity, Temp, and VPD

Previous Results from second.fitting.pdf

- Values clearly co-vary.
- Should consider using
 - Vapor Pressure Deficit
 - * Uses temp to calculate max humidity and then looks at difference with relative humidity.

- Heat Index
 - * Adjusted for birds if available
 - * Use just first terms ($c_1 + c_2 T + c_3 R + c_4 T R + ...$)

Using VPD as Predictor

• In third.fitting we used VPD, but Liz suggested moving back to temp.

Compare Temp, Humidity, and VPD as predictors

• see third.fitting.Rmd

Result

• VPD and temp are highly correlated

Model Fitting

Plots

Formal Model Fits to song_count

Initial

Using

- $x = temp_mean -45C$
- male as a factor
- Filter data

```
tem_ref <- 45
verbose <- 0
trace <- FALSE

## Try filtering the data a bit more
## Goal is to get good starting values

data <- data_ind %>%
    mutate(x1 = (temp_mean - temp_ref)) %>%
    filter( !(male %in% c("T231", "T260")))
```

```
## Error in 'mutate()':
## ! Problem while computing 'x1 = (temp_mean - temp_ref)'.
## Caused by error:
## ! object 'temp_ref' not found
```

```
make_plot = FALSE;
glm poisson <- glm(song count ~</pre>
                      1 + \text{male} + (x1) + I(x1^2), \# + offset(log(count_total_round)),
                   data = data,
                   family = poisson(link = "log")
                   )
## Error in eval(predvars, data, env): object 'x1' not found
summary(glm_poisson)
## Error in h(simpleError(msg, call)): error in evaluating the argument 'object' in selecting a method
if(make_plot) plot(glm_poisson, ask = FALSE)
glm_qpoisson <- update(glm_poisson,</pre>
                       family = quasipoisson(link = "log")
## Error in h(simpleError(msg, call)): error in evaluating the argument 'object' in selecting a method
summary(glm_qpoisson)
## Error in h(simpleError(msg, call)): error in evaluating the argument 'object' in selecting a method
if(make_plot) {
   plot(glm_qpoisson, ask = FALSE)
## Clearly the data is over dispersed
## quasipoisson() doesn't seem to exist for glmer (but likely exists in nlme)
## Try using negative binomial
## This model formulation seems to be correct and converges!!
## HOwever, it ignores the correlation between x1 and I(x1^2)
formula_quad_RE <- song_count \sim -1 + male + x1 + I(x1^2) + (-1 + x1 + I(x1^2)||male)
## This model doesn't converge due to high correlation mentioned above.
formula_quad_RE <- song_count ~ -1 + male + x1 + I(x1^2) + (-1 + x1 + I(x1^2) | male)
## Remove previous results to avoid confusion if model fails to fit
rm(glmer_poisson)
```

Warning in rm(glmer_poisson): object 'glmer_poisson' not found

```
glmer_poisson <- glmer(formula_quad_RE,</pre>
                       data = data,
                       control = glmerControl(
                           optCtrl = list(maxiter = 1E5,
                                          maxfun = 2E6,
                                          trace = trace),
                           optimizer="nloptwrap"),
                       family = poisson(link = "log"),
                       verbose = verbose
## Error in eval(predvars, data, env): object 'x1' not found
summary(glmer_poisson)
## Error in h(simpleError(msg, call)): error in evaluating the argument 'object' in selecting a method
if(make_plot) plot(glmer_poisson, ask = FALSE)
## Results indicate that var in x1^2 term is very small (1E-4)
## So we should drop that.
## Except the model complains about convergence
## Fit using `| male` instead of `|| male` to see what happens
## An extra term, the correlation between the RE terms is estimated.
## Random effects:
## Groups Name Variance Std.Dev. Corr
## male x1
                  0.0472762 0.21743
           I(x1^2) 0.0001374 0.01172 0.96
## Fit improves, but I get convergence complaints.
## This is likely due to the high correlation between the two RE.
## WOudl like to post a question about this to SE
## data_example <- data %>% mutate(f = as.factor(as.integer(male))) %>% select(f, x = temp_mean, y = so.
                                        #tmp <- update(glmer_poisson, formula = formula_quad_RE <- song
glmer_nb <-</pre>
   glmer.nb(formula_quad_RE,
             data = data,
             ## control values are used by the initial optimization
             ## using a poisson glmer model, which doesn't converge
             control = glmerControl(
                 boundary.tol = 0,
                 tolPwrss=1e-1,
                 optCtrl = list(maxiter = 1E5,
                                maxfun = 2E6,
                                trace = trace),
                 optimizer="nloptwrap"),
             ## nb.control values are used by the second optimizer
             ## Note need to set own optCtrl values
```

```
nb.control = list(
                 optCtrl = list(maxit = 1000,
                                maxfun = 2E5)),
             verbose = verbose
## Error in eval(predvars, data, env): object 'x1' not found
## Try dropping the quadratic RE
formula_linear_RE <- song_count \sim -1 + male + x1 + I(x1^2) + (-1 + x1| male)
rm(glmer_poisson_linear)
## Warning in rm(glmer_poisson_linear): object 'glmer_poisson_linear' not found
try(glmer_poisson_linear <- update(glmer_poisson, formula = formula_linear_RE,
                                   verbose = verbose)
   );
## Error in h(simpleError(msg, call)) :
    error in evaluating the argument 'object' in selecting a method for function 'update': object 'glm
if(exists("glmer poisson linear")){
    summary(glmer_poisson_linear);
    if(make_plot) plot(glmer_poisson_linear, ask = FALSE)
}
## This model fails to converge, strangely.
## NB model fails to converge
## This appears to be because the variances of the RE become very small (on the order of E-11!)
## See B. Bolker's notes on GLMM fitting and comparisons
## https://bbolker.github.io/mixedmodels-misc/ecostats chap.html#diagnostics-and-model-summaries
glmer_nb_linear <-</pre>
   glmer.nb(formula_linear_RE,
             data = data,
             ## control values are used by the initial optimization
             ## using a poisson glmer model, which doesn't converge
             control = glmerControl(
                 boundary.tol = 0,
                 tolPwrss=1e-1,
                 optCtrl = list(maxiter = 1E5,
                                maxfun = 2E6,
                                trace = trace),
                 optimizer="nloptwrap"),
             ## nb.control values are used by the second optimizer
             ## Note need to set own optCtrl values
             nb.control = list(
                 optCtrl = list(maxit = 1000,
                                maxfun = 2E5)),
             verbose = verbose
```

```
## Error in eval(predvars, data, env): object 'x1' not found
rm(make_plot, formula_quad_RE, formula_linear_RE)
```

Result

- Model doesn't converge.
- Output includes > Model is nearly unidentifiable: very large eigenvalue > Rescale variables?;Model is nearly unidentifiable: large eigenvalue ratio > Rescale variables? To me this initially suggested we should use song_prop and the glmer weights function. However, I've made some progress in my understanding of ME models.

Use All Round 2 Data

select(fit_db, is.na(fit[[1]]))

Create fit_db

```
vec_{temp_ref} \leftarrow c(27, 35, 44, 45)
## Create a tibble for holding model result
## filter out rows that won't get defined.
fit_db <- expand_grid(temp_ref = vec_temp_ref, model = c("glm", "glmer"), re = c("none", "linear", "qua
   filter( !(model == "glm" & est_cov ==TRUE)) %>%
   filter( !(model == "glmer" & dist =="qpoisson")) %>%
    add_column(fit = list(list(NA)) )
## Try to understand how to work with fit_db
## Filter fit's that are NA, but doesn't work if I assign glm objects to fit[[i]]
subset(fit_db, is.na(fit[[1]]))
## # A tibble: 112 x 6
##
      temp_ref model re
                            est_cov dist
                                             fit
##
         <dbl> <chr> <chr>
                            <lgl>
                                    <chr>>
                                             t>
                                    poisson <list [1]>
##
                            FALSE
   1
            27 glm
                     none
                            FALSE
##
   2
            27 glm
                     none
                                    qpoisson <list [1]>
##
   3
            27 glm
                     none
                            FALSE
                                    nb
                                             t [1]>
##
   4
            27 glm
                     linear FALSE
                                    poisson <list [1]>
##
  5
            27 glm
                     linear FALSE
                                    qpoisson <list [1]>
##
  6
            27 glm
                     linear FALSE
                                             t [1]>
                                    nb
## 7
            27 glm
                     quad
                            FALSE
                                    poisson <list [1]>
## 8
                            FALSE
                                    qpoisson <list [1]>
            27 glm
                     quad
## 9
            27 glm
                            FALSE
                                             <list [1]>
                     quad
## 10
            27 glm
                            FALSE
                                    poisson <list [1]>
                     both
## # ... with 102 more rows
## Doesn't know what fit is
```

```
## Error in 'select()':
## ! Problem while evaluating 'is.na(fit[[1]])'.
## Caused by error:
## ! object 'fit' not found
```

GLM Fits

• Just FE

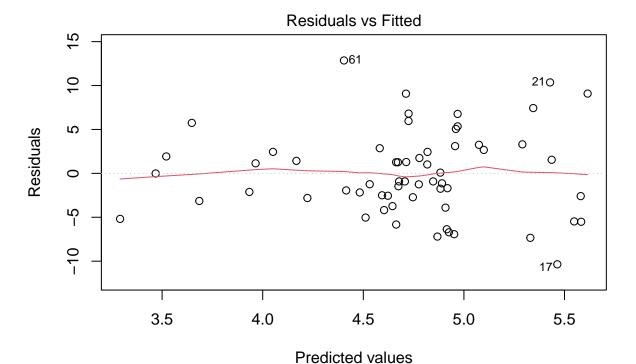
Fit Fixed Effects models

```
### Fit Fixed Effects models
est_cov <- FALSE
verbose <- 0
trace <- FALSE
for(temp_ref in vec_temp_ref ){
    data <- data_ind %>%
        mutate(x1 = (temp_mean - temp_ref))
    print(paste("temp_ref:", temp_ref))
    re <- "none"
    glm_poisson <-</pre>
        glm(song\_count \sim 1 + male + (x1) + I(x1^2),
            data = data,
            family = poisson(link = "log")
    summary(glm_poisson)
    plot(glm_poisson, ask = FALSE)
    fit_db[fit_db$temp_ref==temp_ref & fit_db$model == "glm" & fit_db$dist == "poisson" & fit_db$re =
    glm_qpoisson <- update(glm_poisson,</pre>
                           family = quasipoisson(link = "log")
    summary(glm_qpoisson)
    plot(glm_qpoisson, ask = FALSE)
    fit_db[fit_db$temp_ref==temp_ref & fit_db$model == "glm" & fit_db$dist == "qpoisson" & fit_db$re
    glm_qpoisson <- update(glm_poisson,</pre>
                           family = quasipoisson(link = "log")
    summary(glm_qpoisson)
    ## Note the dispersion parameter is 29.7 rather than 1!!
    ## Clearly the data is over dispersed
```

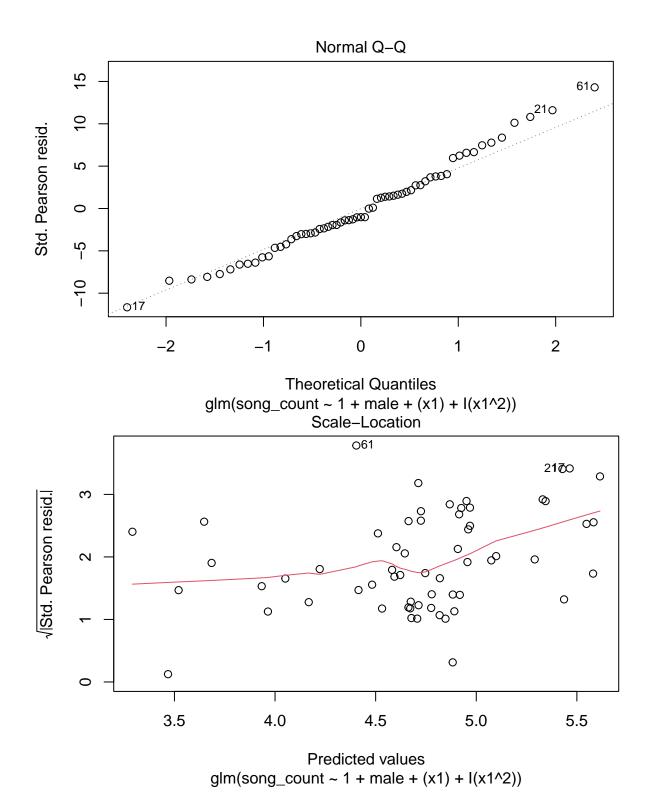
```
## quasipoisson() doesn't seem to exist for glmer (but likely exists in nlme)
plot(glm_qpoisson, ask = FALSE)

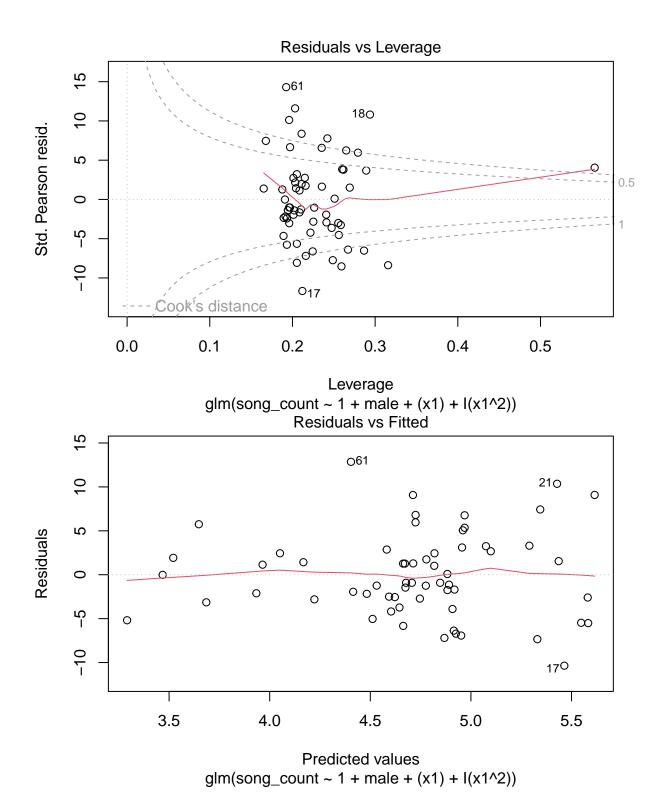
fit_db[fit_db$temp_ref==temp_ref & fit_db$model == "glm" & fit_db$dist == "qpoisson" & fit_db$re == "glm" of the fit_db$dist == "qpoisson" of the fit_db$re == "glm" of the fit_db$dist == "qpoisson" of the fit_db$re == "glm" of the fit_db$dist == "qpoisson" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$dist == "nb" of the fit_db$re == re == "glm" of the fit_db$fit_db$re == re == "glm" of the fit_db$fit_db$re == re == "glm" of the fit_db$fit_db$fit_db$re == re == "glm" of the fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fit_db$fi
```

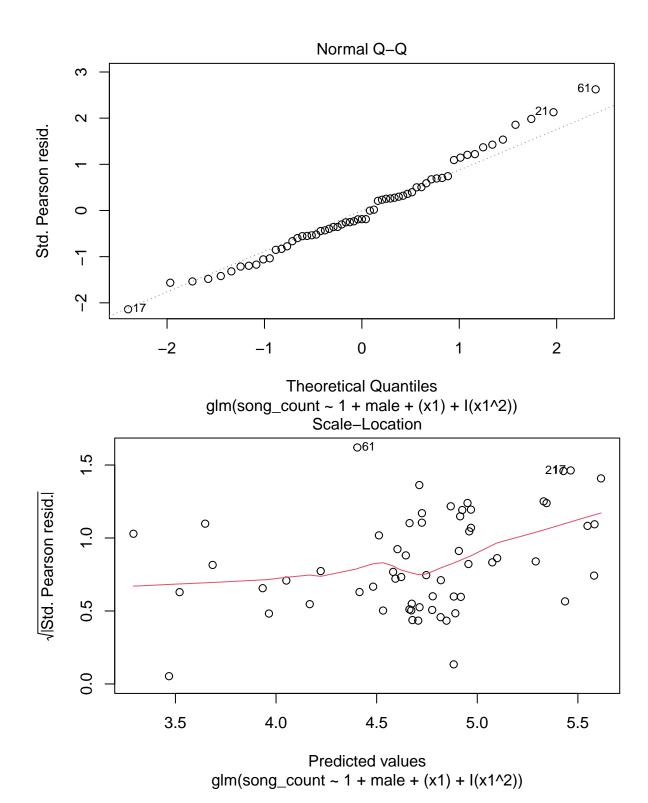
[1] "temp_ref: 27"

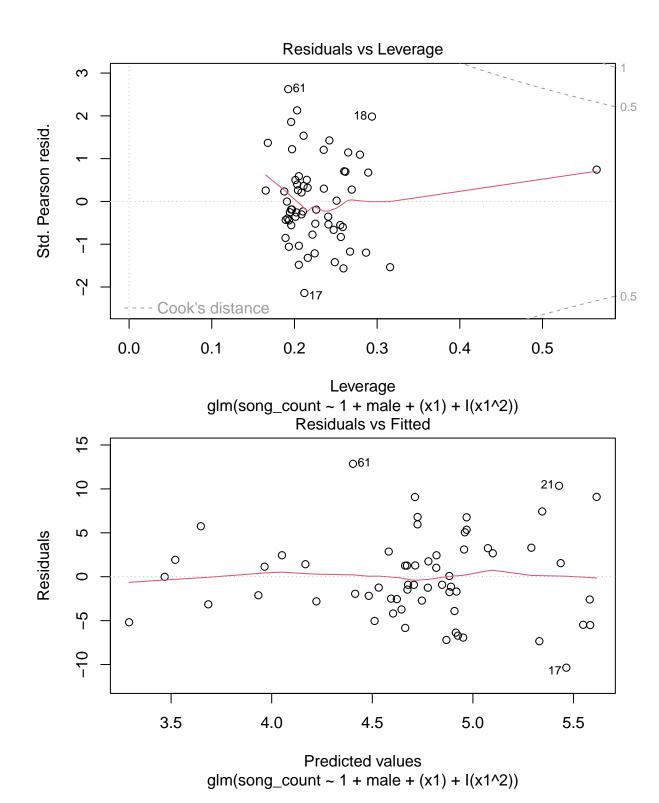


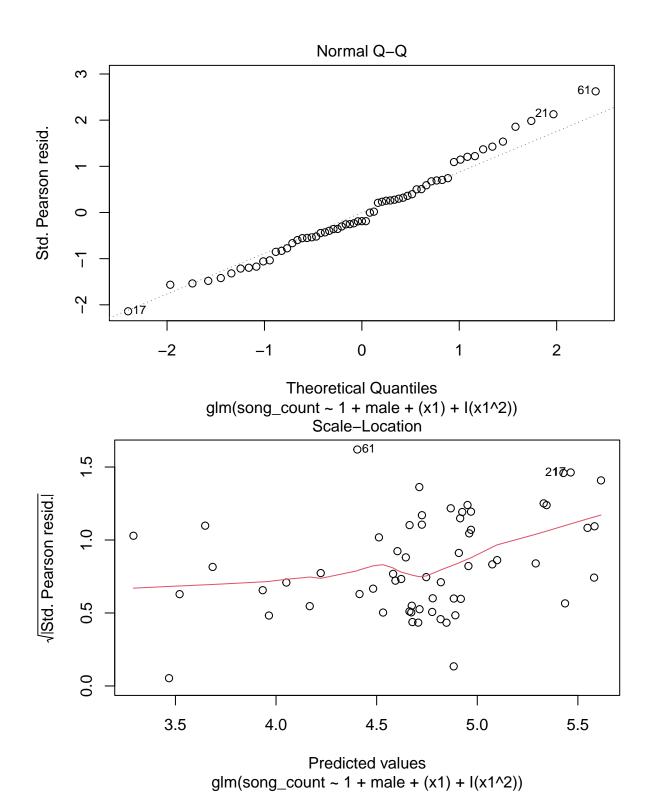
 $glm(song_count \sim 1 + male + (x1) + I(x1^2))$

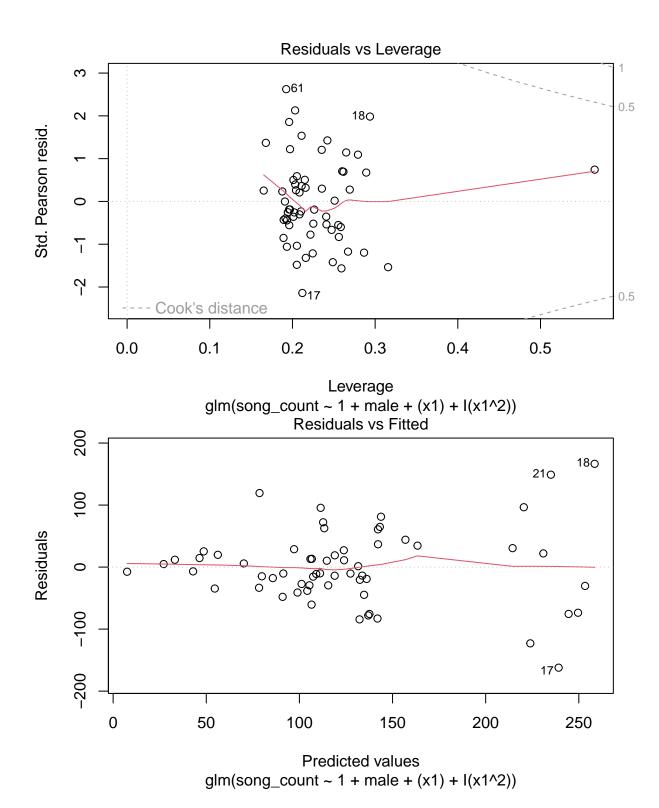


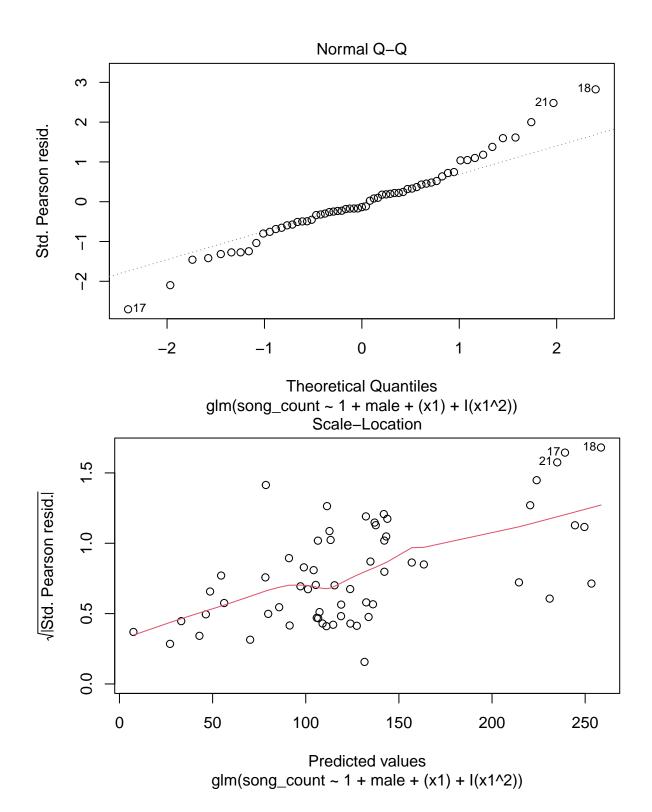


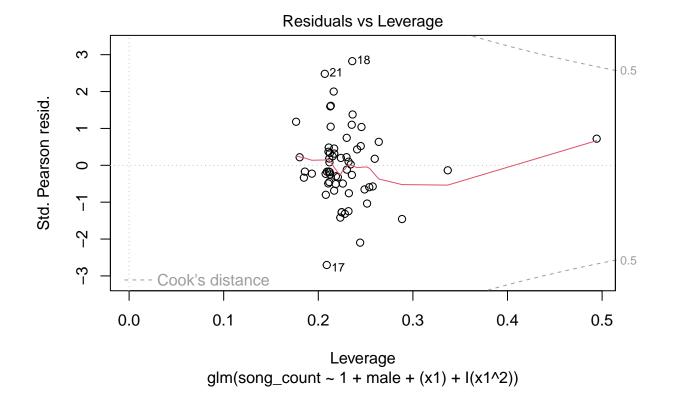




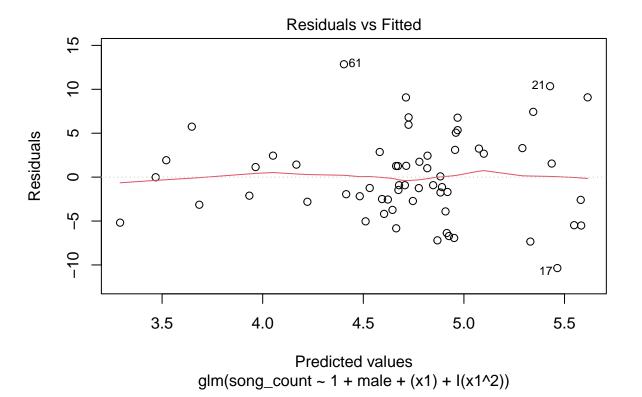


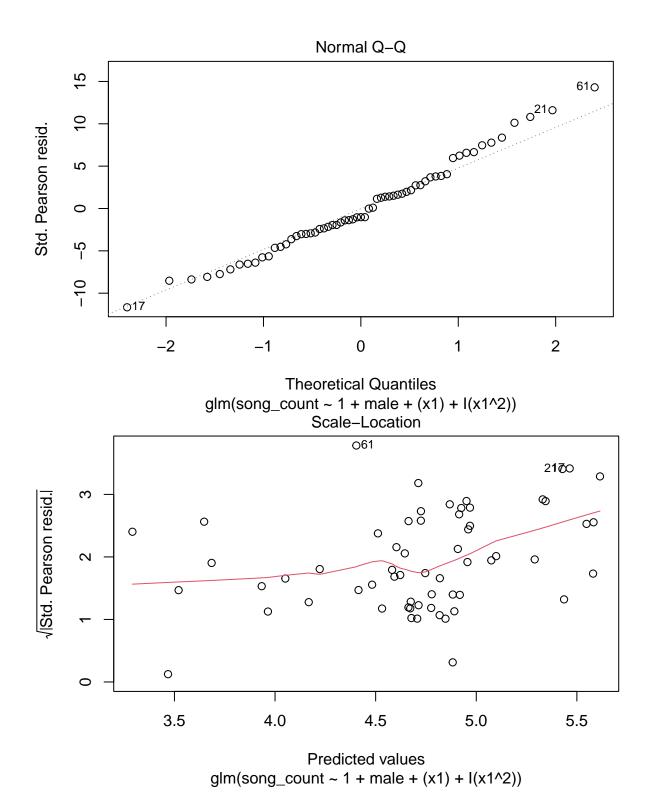


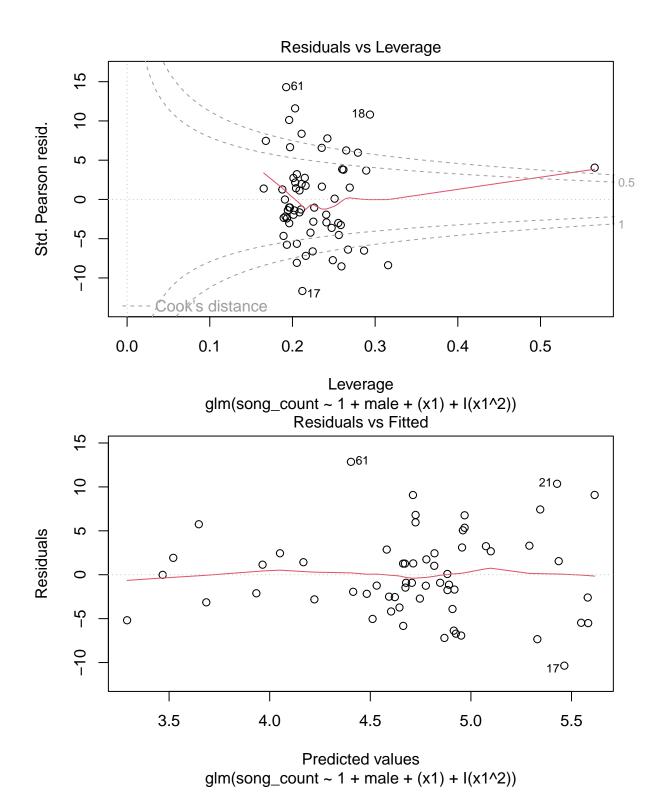


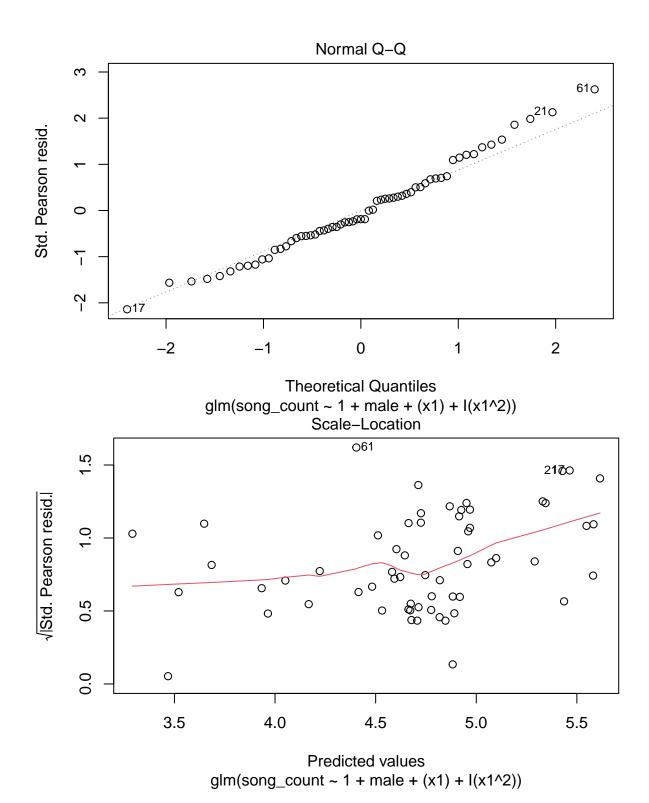


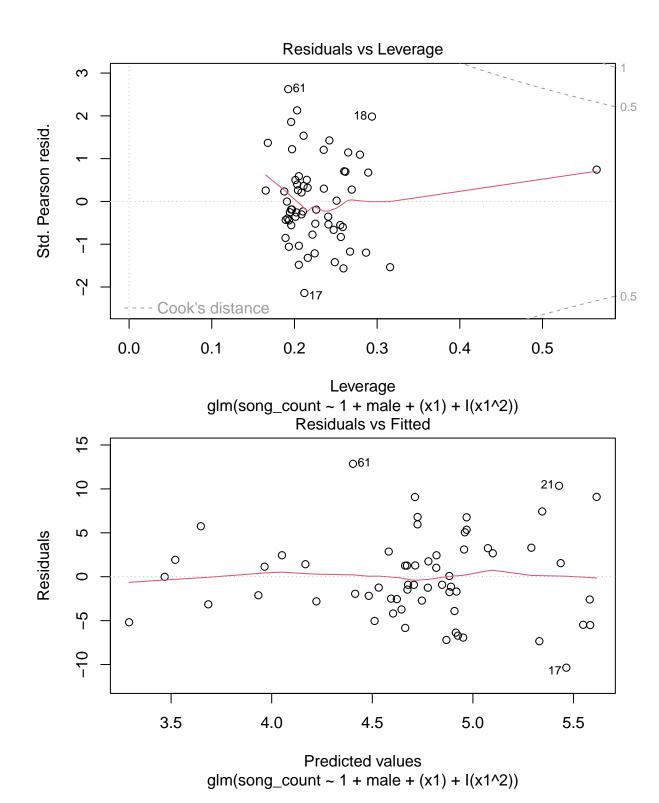
[1] "temp_ref: 35"

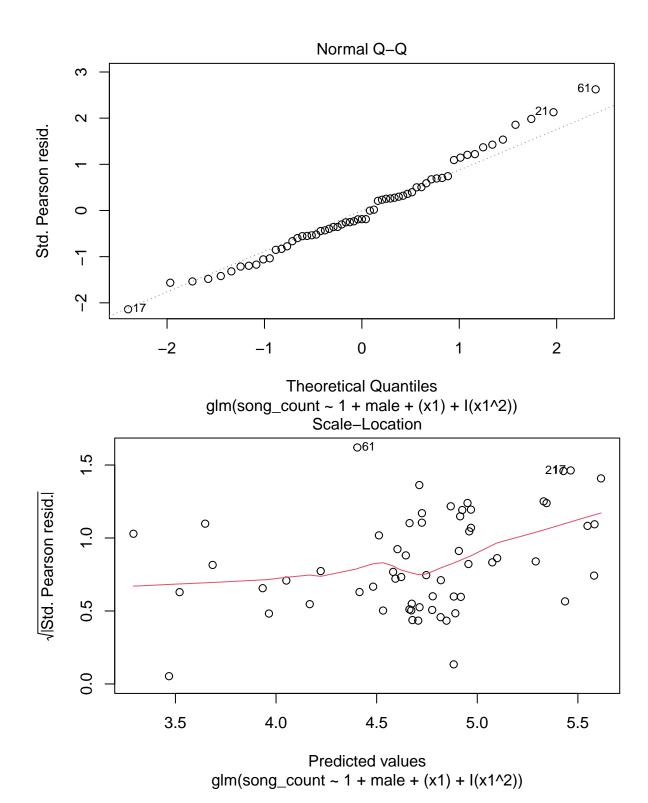


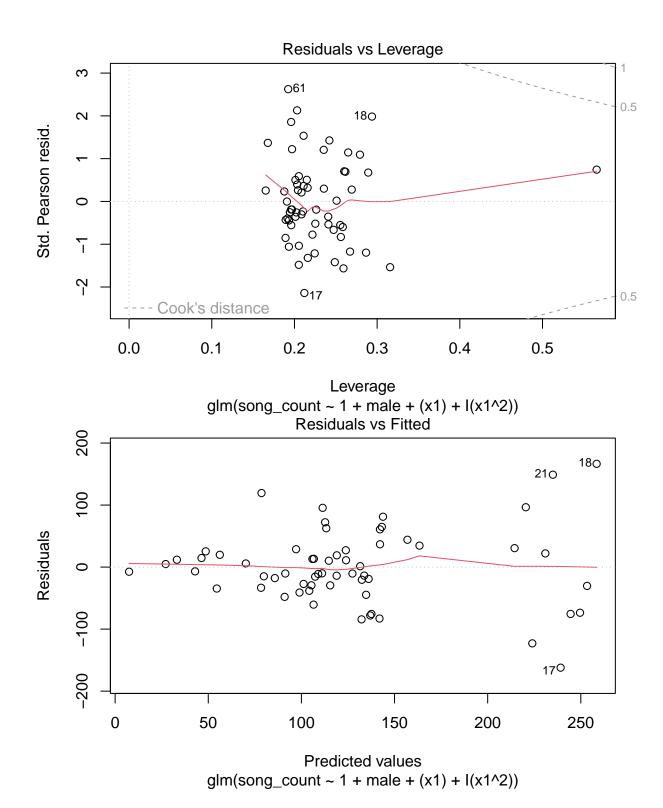


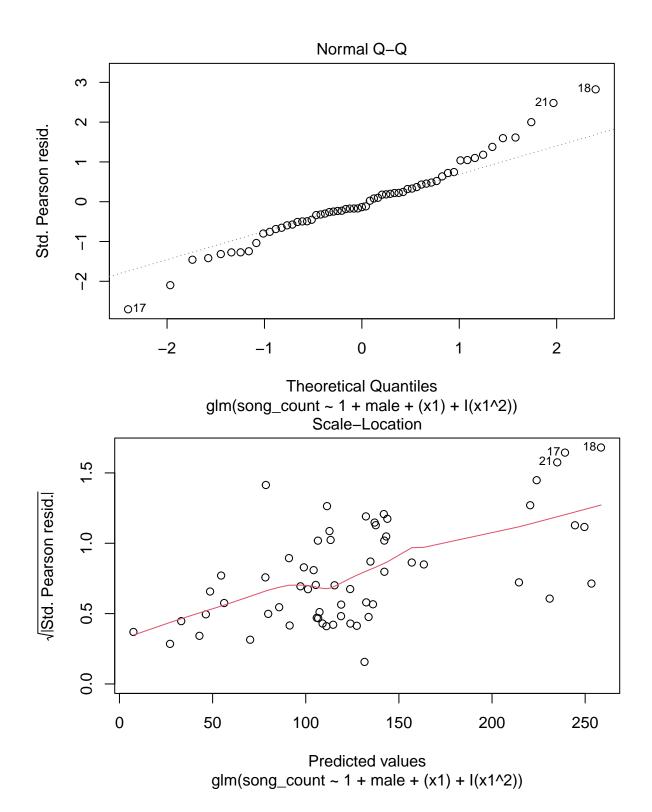


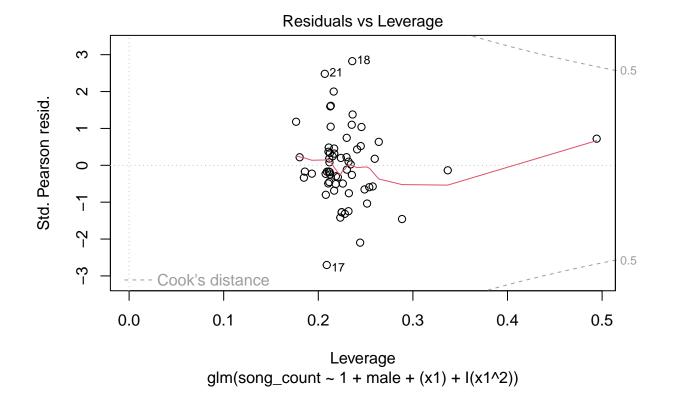




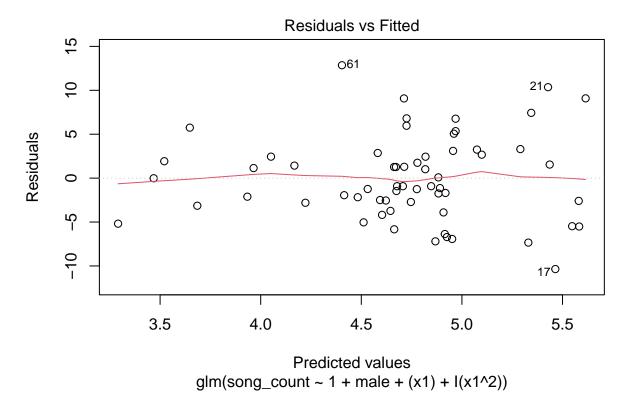


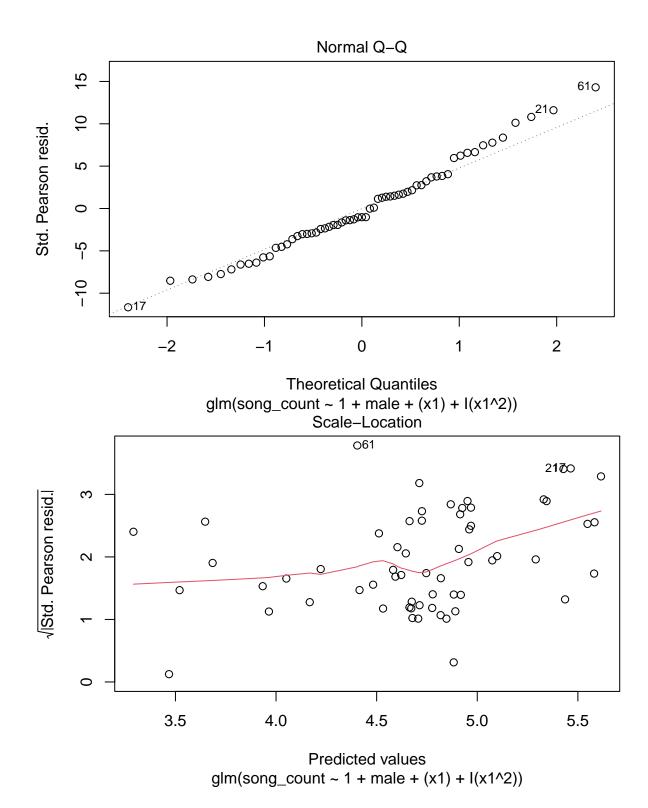


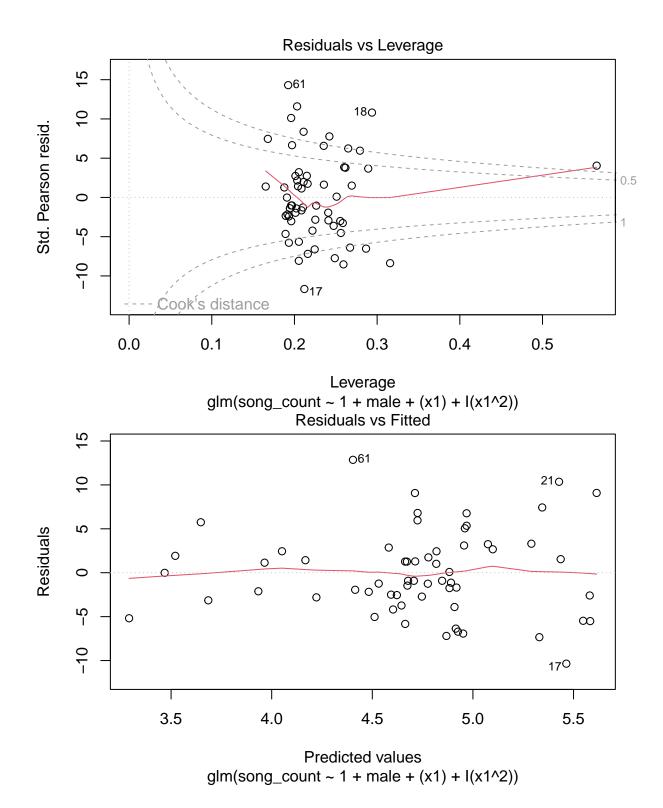


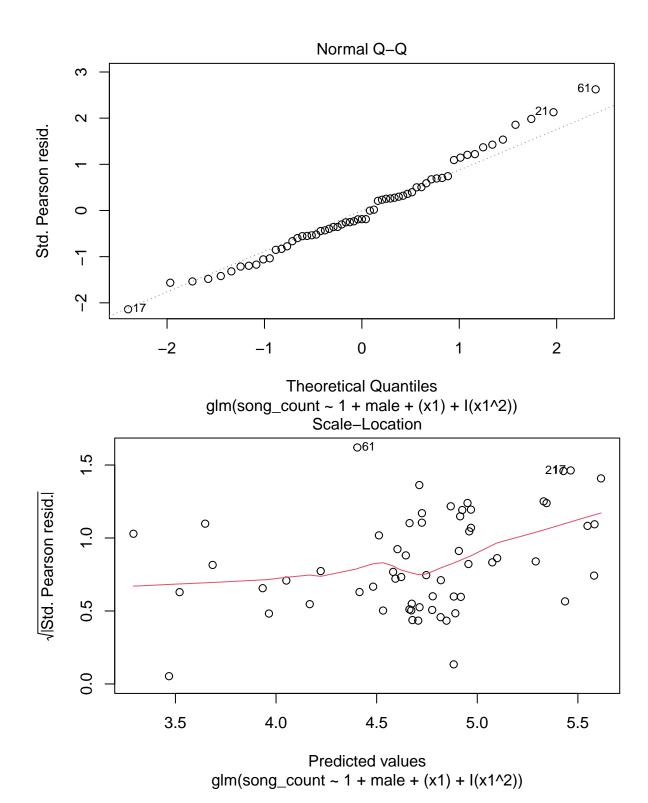


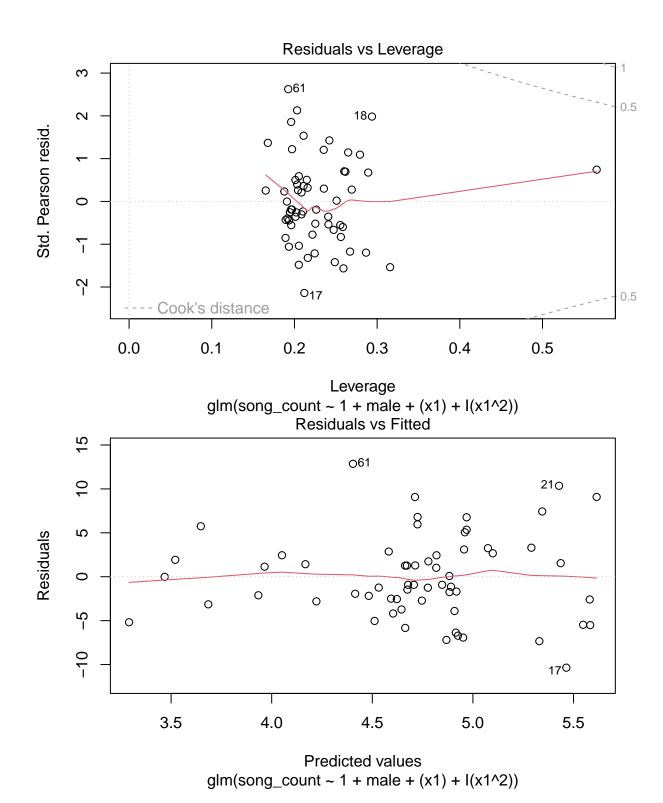
[1] "temp_ref: 44"

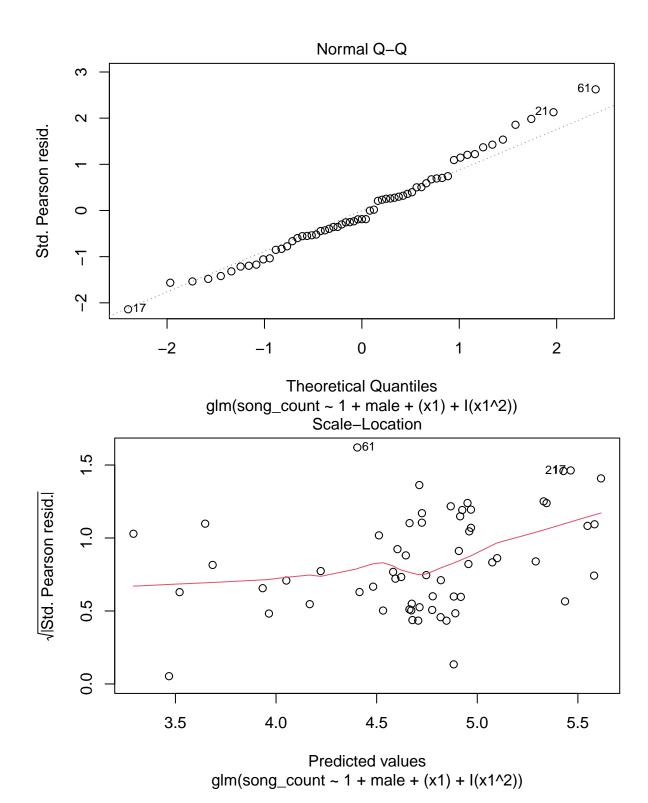


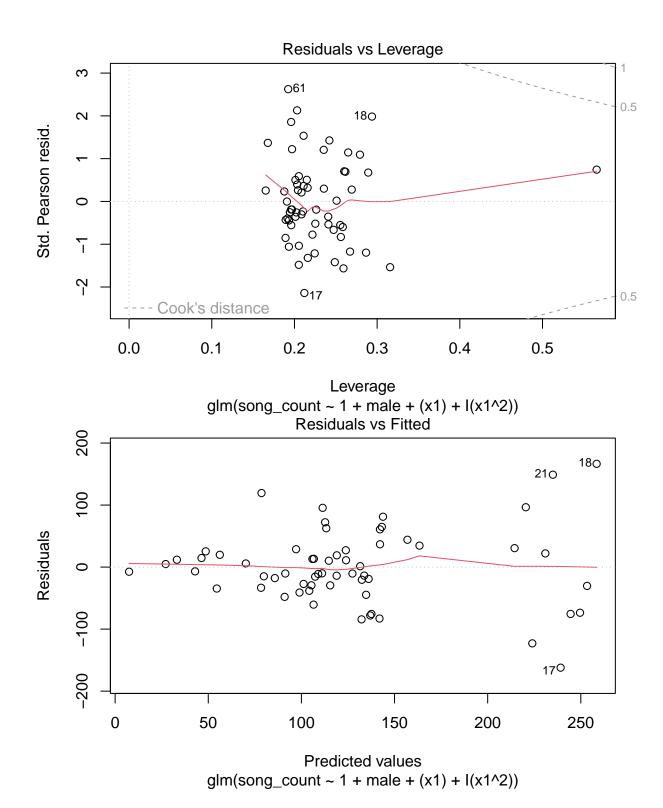


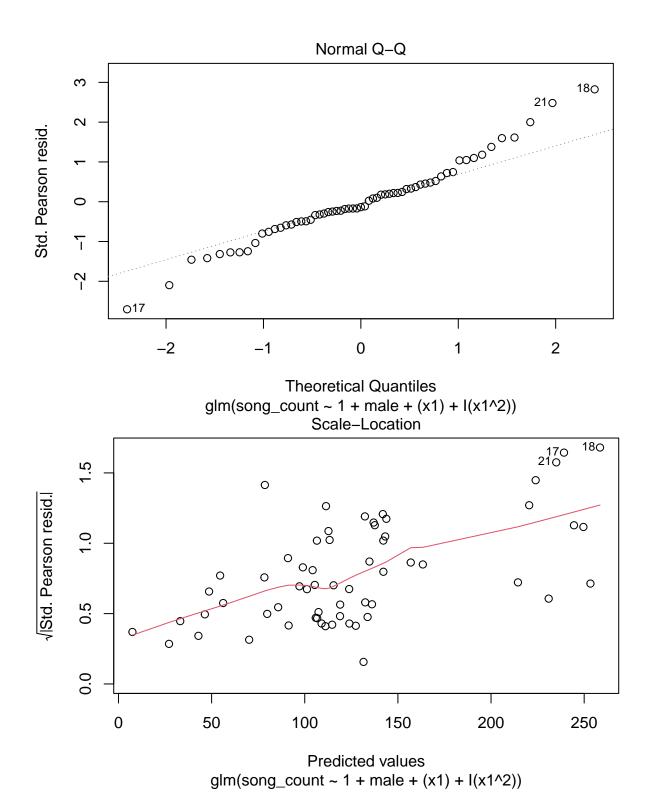


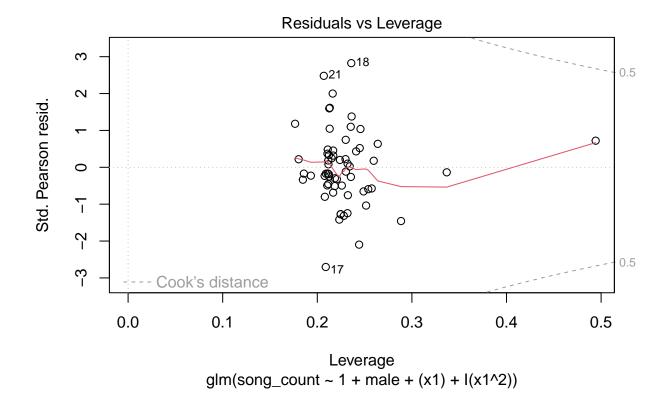




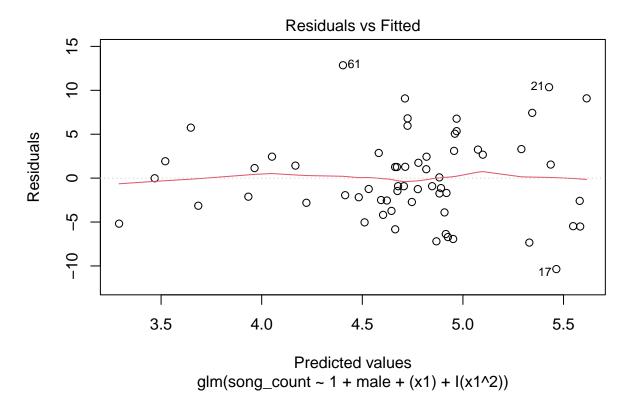


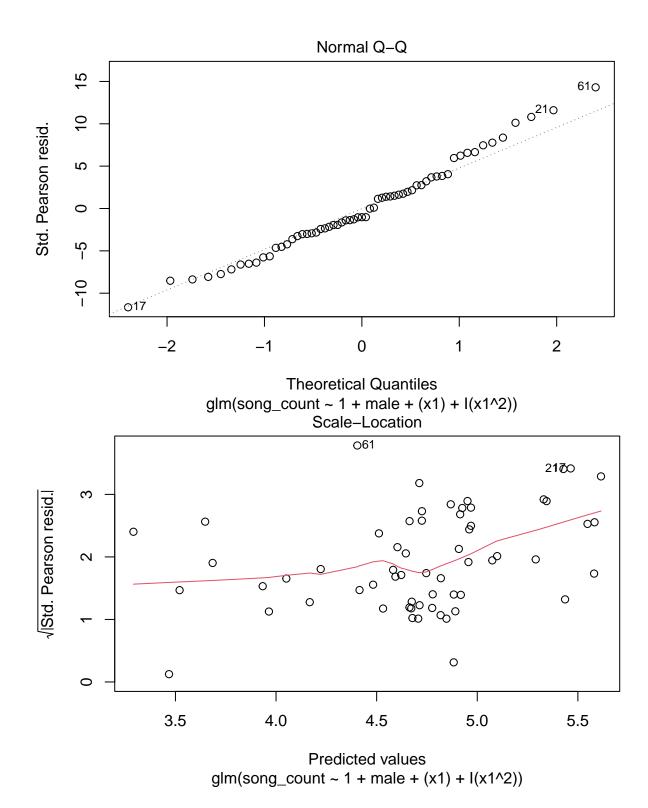


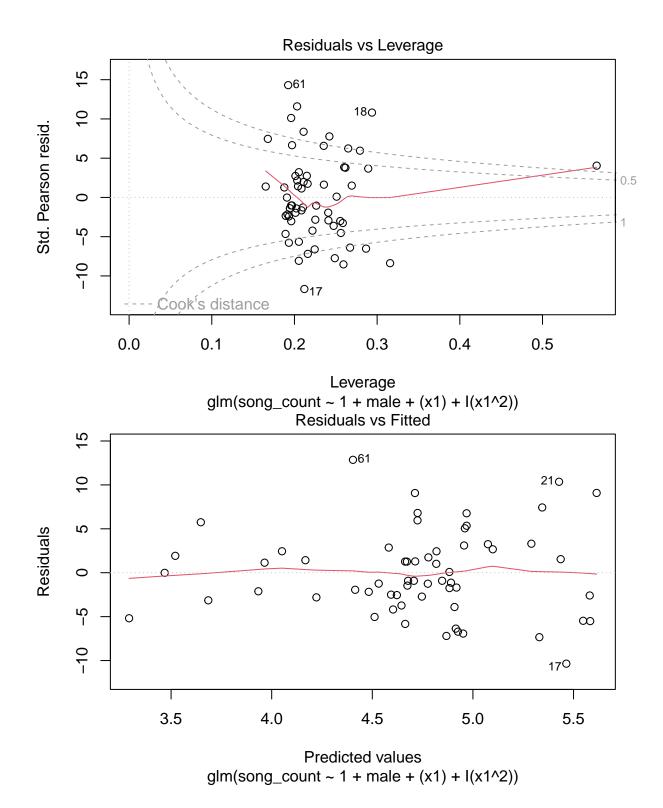


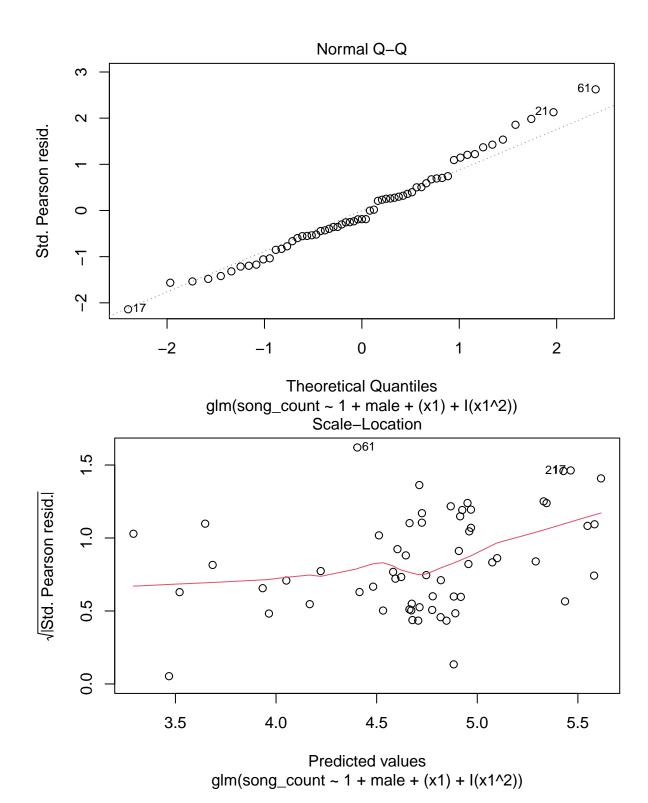


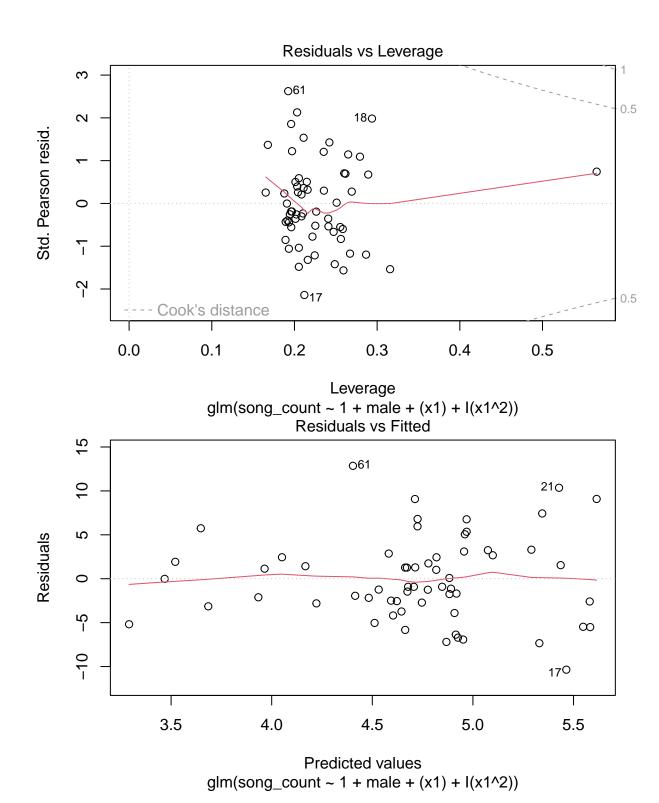
[1] "temp_ref: 45"

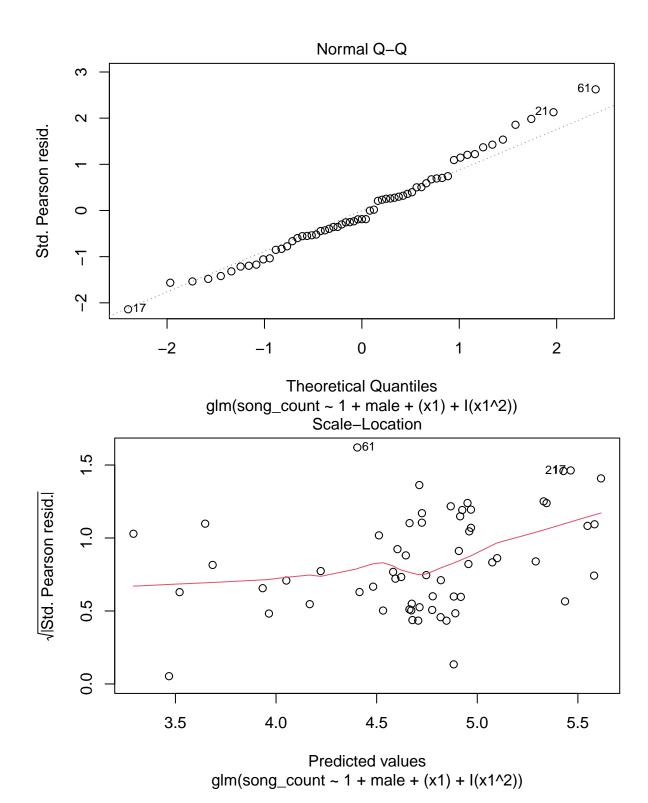


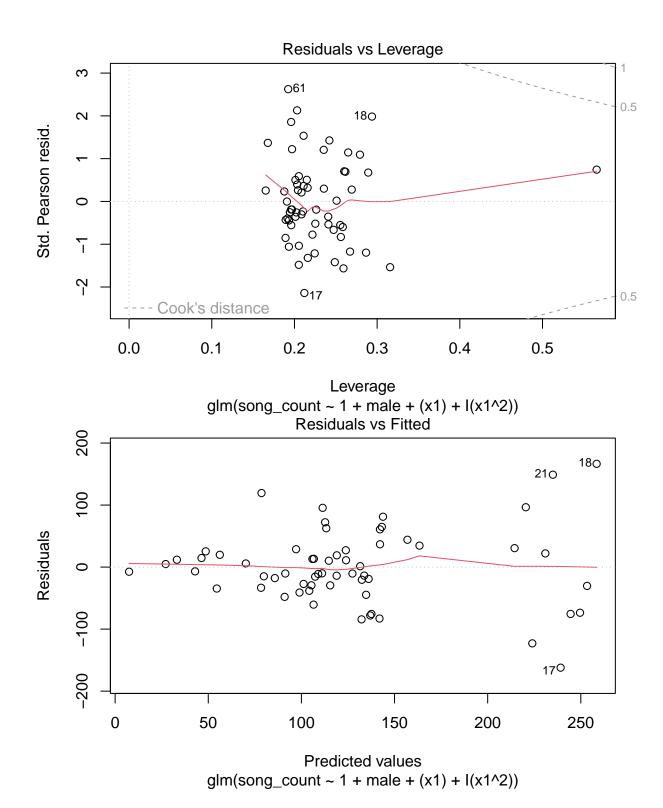


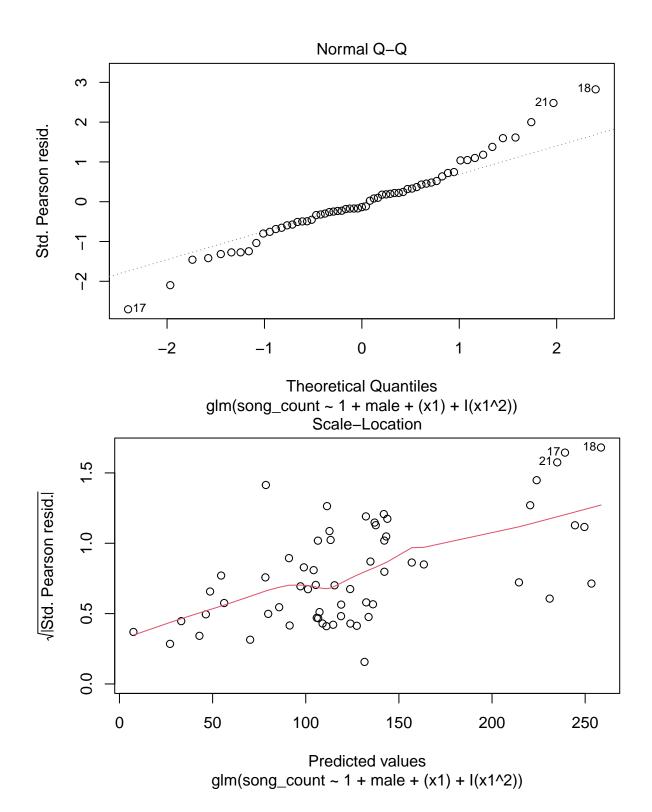


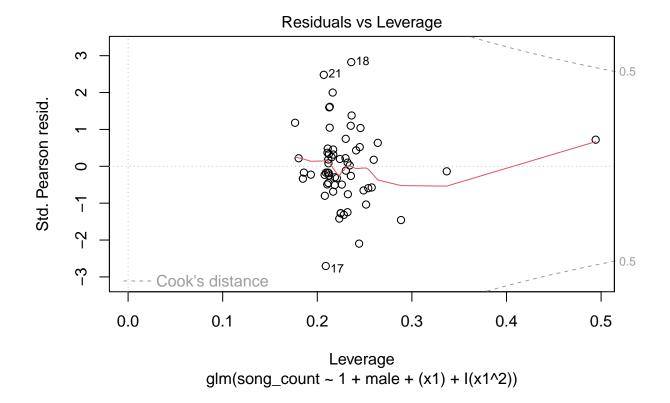












- glm of poisson gives significant x1 and x1^2 terms, but if I adjust for overdispersion using qpoisson they become insignicant.
- I see the same result, enormous CI for x1 and $x1^2$ terms, with glm_nb .
- These results are the motivation for using RE. That is, to deal with the heterogeneity in a more controlled manner. However, none of the glmer_nb model fittings work well
 - There appears to be a negative correlation between the RE for x1 and x1^2.
 - The variance in the RE are pretty small which makes me wonder why don't the fixed effect models work well enough.

GLMM Fits

No RE co-var: || Formulation

- On reflection the "both" is the only RE setting where $\mid \mid$ should differ from \mid

```
est_cov <- FALSE
verbose <- 0
trace <- FALSE
optimizer <- "bobyqa" #"nloptwrap" # "bobyqa"

## Fit random effects models
for(dist in c("poisson", "nb")) {
    for(temp_ref in c(35)) { # vec_temp_ref ){
        for(re in c("linear", "quad", "both")) {
}</pre>
```

```
switch(re,
       both = (formula_re <- song_count \sim -1 + male + x1 + I(x1^2) + (-1 + x1 + I(x1^2) |
       linear = (formula_re <- song_count \sim -1 + male + x1 + I(x1^2) + (-1 + x1 || male))
       quad = (formula_re <- song_count \sim -1 + male + \times 1 + I(\times 1^2) + (-1 + I(\times 1^2) || male
## This model formulation seems to be correct and converges with optimizer = "bobyqa"!!
## HOwever, it ignores the correlation between x1 and I(x1^2)
## formula quad RE <- song count \sim -1 + \text{male} + x1 + I(x1^2) + (-1 + x1 + I(x1^2)) / \text{male}
## I'm not sure what that means
if(dist == "poisson"){
    print(paste("temp_ref =", temp_ref,"glmer_poisson", "re = ", re))
    glmer_poisson <- glmer(formula = formula_re,</pre>
                            data = data,
                            control = glmerControl(
                                optCtrl = list(maxiter = 1E5,
                                                maxfun = 2E6,
                                                trace = trace),
                                optimizer=optimizer),
                            family = poisson(link = "log"),
                            verbose = verbose
                            )
    print(summary(glmer_poisson))
    plot(glmer_poisson, ask=FALSE)
    fit_db[fit_db$temp_ref==temp_ref & fit_db$model == "glmer" & fit_db$dist == "poisson"
    print("evaluate overdispersion")
    glmer_qpoisson <- overdisp.glmer(glmer_poisson)</pre>
}
if(dist == "nb"){
    print(paste("temp_ref =", temp_ref,"glmer_nb", "re = ", re))
    try(glmer_nb <-</pre>
            glmer.nb(formula = formula_re,
                     data = data,
                      ## control values are used by the initial optimization
                      ## using a poisson glmer model, which doesn't converge
                      control = glmerControl(
                         boundary.tol = 0,
                          tolPwrss=1e-1,
                          optCtrl = list(maxiter = 1E5,
                                         maxfun = 2E6,
                                         trace = trace),
                          optimizer = optimizer),
                      ## nb.control values are used by the second optimizer
                      ## Note need to set own optCtrl values
                      nb.control = list(
                          optCtrl = list(maxit = 1000,
                                         \max fun = 2E5)),
                      verbose = verbose
```

```
)
                plot(glmer_nb, ask = FALSE)
                print(summary(glmer_nb))
                fit_db[fit_db$temp_ref==temp_ref & fit_db$model == "glmer" & fit_db$dist == "nb" & fi
           }
       }
   }
}
## [1] "temp_ref = 35 glmer_poisson re = linear"
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.015591 (tol = 0.002, component 1)
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unide
## - Rescale variables?
## Generalized linear mixed model fit by maximum likelihood (Laplace
    Approximation) [glmerMod]
## Family: poisson ( log )
## Formula: song_count \sim -1 + male + x1 + I(x1^2) + (-1 + x1 || male)
##
     Data: data
## Control: glmerControl(optCtrl = list(maxiter = 1e+05, maxfun = 2e+06,
      trace = trace), optimizer = optimizer)
##
##
##
        AIC
                       logLik deviance df.resid
                BIC
                      -700.8
##
     1431.7
             1463.3
                               1401.7
##
## Scaled residuals:
      Min
               1Q Median
                                3Q
## -9.4099 -2.3601 -0.3221 1.7150 14.1120
##
## Random effects:
## Groups Name Variance Std.Dev.
              0.002995 0.05473
         x1
## Number of obs: 61, groups: male, 12
##
## Fixed effects:
##
            Estimate Std. Error z value Pr(>|z|)
## maleT229 3.958900
                       0.091330 43.347 < 2e-16 ***
## maleT231 4.240132
                       0.089705 47.267 < 2e-16 ***
## maleT234 3.758839 0.115413 32.569 < 2e-16 ***
```

maleT235 4.504916 0.082918 54.330 < 2e-16 ***

```
## maleT236 3.876809
                       0.112270 34.531 < 2e-16 ***
## maleT237 4.058857
                       0.101564 39.964 < 2e-16 ***
                       0.081779 61.077 < 2e-16 ***
## maleT243 4.994830
                       0.121522 31.698 < 2e-16 ***
## maleT244 3.852043
## maleT247 4.406083 0.089316 49.331 < 2e-16 ***
## maleT257 5.526669 0.063079 87.615 < 2e-16 ***
## maleT258 2.833279
                       0.156104 18.150 < 2e-16 ***
                       0.113902 41.222 < 2e-16 ***
## maleT260 4.695259
## x1
           -0.118092
                       0.021028 -5.616 1.95e-08 ***
## I(x1^2) -0.005618 0.000760 -7.393 1.44e-13 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Correlation matrix not shown by default, as p = 14 > 12.
## Use print(summary(glmer_poisson), correlation=TRUE) or
      vcov(summary(glmer_poisson))
                                          if you need it
## optimizer (bobyqa) convergence code: 0 (OK)
## Model failed to converge with max|grad| = 0.015591 (tol = 0.002, component 1)
## Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
## unused control arguments ignored
## [1] "evaluate overdispersion"
## Residual deviance: 939.335 on 46 degrees of freedom (ratio: 20.42)
## [1] "temp_ref = 35 glmer_poisson re = quad"
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.00649726 (tol = 0.002, component 1)
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unide:
## - Rescale variables?
## Generalized linear mixed model fit by maximum likelihood (Laplace
##
    Approximation) [glmerMod]
## Family: poisson (log)
## Formula: song_count \sim -1 + \text{male} + x1 + I(x1^2) + (-1 + I(x1^2) || male)
##
     Data: data
## Control: glmerControl(optCtrl = list(maxiter = 1e+05, maxfun = 2e+06,
      trace = trace), optimizer = optimizer)
##
##
##
       AIC
                BIC
                      logLik deviance df.resid
##
    1451.5
             1483.1
                     -710.7 1421.5
##
## Scaled residuals:
```

```
10 Median
                              3Q
## -8.9198 -2.4339 -0.2126 1.5026 14.6714
##
## Random effects:
## Groups Name
                 Variance Std.Dev.
## male I(x1^2) 8.531e-06 0.002921
## Number of obs: 61, groups: male, 12
## Fixed effects:
##
            Estimate Std. Error z value Pr(>|z|)
## maleT229 4.124778 0.074827 55.124 < 2e-16 ***
## maleT231 4.266849 0.074674 57.140 < 2e-16 ***
## maleT234 3.743742 0.090030 41.583 < 2e-16 ***
## maleT235 4.692110 0.069463 67.549 < 2e-16 ***
## maleT236 4.230641 0.083474 50.682 < 2e-16 ***
## maleT237 4.179025 0.080902 51.656 < 2e-16 ***
## maleT243 4.726213 0.067627 69.886 < 2e-16 ***
## maleT244 3.976604 0.090476 43.952 < 2e-16 ***
## maleT247 4.401292 0.073165 60.155 < 2e-16 ***
## maleT257 5.311227 0.056622 93.802 < 2e-16 ***
## maleT258 3.084881 0.117646 26.222 < 2e-16 ***
## maleT260 4.344404 0.084856 51.197 < 2e-16 ***
          ## x1
## I(x1^2) -0.005631 0.001153 -4.884 1.04e-06 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Correlation matrix not shown by default, as p = 14 > 12.
## Use print(summary(glmer_poisson), correlation=TRUE) or
      vcov(summary(glmer_poisson))
                                        if you need it
## optimizer (bobyqa) convergence code: 0 (OK)
## Model failed to converge with max|grad| = 0.00649726 (tol = 0.002, component 1)
## Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
## unused control arguments ignored
## [1] "evaluate overdispersion"
## Residual deviance: 968.391 on 46 degrees of freedom (ratio: 21.052)
## [1] "temp_ref = 35 glmer_poisson re = both"
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Generalized linear mixed model fit by maximum likelihood (Laplace
    Approximation) [glmerMod]
## Family: poisson ( log )
## Formula: song count \sim -1 + male + x1 + I(x1^2) + (-1 + x1 + I(x1^2) |
##
      male)
```

```
##
      Data: data
## Control: glmerControl(optCtrl = list(maxiter = 1e+05, maxfun = 2e+06,
##
      trace = trace), optimizer = optimizer)
##
##
       ATC
                BIC
                      logLik deviance df.resid
              1365.0
                      -649.6
                               1299.2
##
     1331.2
##
## Scaled residuals:
##
     Min
             1Q Median
                           3Q
## -8.240 -2.028 -0.278 1.208 15.096
## Random effects:
## Groups Name
                  Variance Std.Dev.
## male
                   0.0782063 0.27965
          x1
## male.1 I(x1^2) 0.0002153 0.01467
## Number of obs: 61, groups: male, 12
##
## Fixed effects:
            Estimate Std. Error z value Pr(>|z|)
##
## maleT229 4.153562
                       0.150360 27.624
                                          <2e-16 ***
## maleT231 4.728800
                     0.146638 32.248
                                          <2e-16 ***
## maleT234 3.281943
                       0.216555 15.155
                                          <2e-16 ***
## maleT235 5.094337
                       0.142709 35.697
                                          <2e-16 ***
## maleT236 3.062421
                       0.189852 16.131
                                          <2e-16 ***
## maleT237 4.375082 0.178590 24.498
                                          <2e-16 ***
## maleT243 5.027424
                       0.147155 34.164
                                          <2e-16 ***
## maleT244 3.849734
                       0.286147 13.454
                                          <2e-16 ***
## maleT247 3.787706
                       0.186288 20.333
                                          <2e-16 ***
## maleT257 5.654247
                       0.111432 50.742
                                          <2e-16 ***
## maleT258 1.010165
                       0.398468
                                 2.535
                                          0.0112 *
                       0.351424 18.704
## maleT260 6.573144
                                          <2e-16 ***
## x1
            -0.114864
                       0.082535 -1.392
                                          0.1640
## I(x1^2) -0.005112
                       0.004342 -1.177
                                          0.2391
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Correlation matrix not shown by default, as p = 14 > 12.
## Use print(summary(glmer_poisson), correlation=TRUE) or
       vcov(summary(glmer_poisson))
                                          if you need it
## optimizer (bobyqa) convergence code: 0 (OK)
## unused control arguments ignored
## [1] "evaluate overdispersion"
## Residual deviance: 728.431 on 45 degrees of freedom (ratio: 16.187)
## [1] "temp_ref = 35 glmer_nb re = linear"
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
```

```
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.063488 (tol = 0.002, component 1)
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unide:
## - Rescale variables?
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in optTheta(g1, interval = interval, tol = tol, verbose = verbose, :
## unused control arguments ignored
## boundary (singular) fit: see help('isSingular')
## Generalized linear mixed model fit by maximum likelihood (Laplace
     Approximation) [glmerMod]
## Family: Negative Binomial(4.895) (log)
## Formula: song_count ~ -1 + male + x1 + I(x1^2) + (-1 + x1 || male)
      Data: data
## Control:
  glmerControl(boundary.tol = 0, tolPwrss = 0.1, optCtrl = list(maxiter = 1e+05,
       maxfun = 2e+06, trace = trace), optimizer = optimizer)
##
##
##
                BIC
                      logLik deviance df.resid
        AIC
##
      680.2
              714.0
                      -324.1
                                 648.2
##
## Scaled residuals:
                1Q Median
## -1.9753 -0.6912 -0.1580 0.5586 3.0699
##
## Random effects:
## Groups Name Variance Std.Dev.
## male x1
              3.432e-11 5.859e-06
## Number of obs: 61, groups: male, 12
##
## Fixed effects:
            Estimate Std. Error z value Pr(>|z|)
## maleT229 4.275715
                       0.301350 14.189
                                          <2e-16 ***
## maleT231 4.301052
                       0.302327 14.227
                                          <2e-16 ***
## maleT234 3.557692
                        0.305281 11.654
                                           <2e-16 ***
                        0.316681 15.498
## maleT235 4.907960
                                           <2e-16 ***
## maleT236 4.432163
                        0.300701 14.739
                                           <2e-16 ***
## maleT237 4.262585
                        0.309520 13.772
                                           <2e-16 ***
## maleT243 4.355303
                        0.289683 15.035
                                           <2e-16 ***
## maleT244 4.002712
                        0.324817 12.323
                                           <2e-16 ***
                        0.305795 13.790
## maleT247 4.216963
                                           <2e-16 ***
## maleT257 4.992306
                        0.291354 17.135
                                           <2e-16 ***
## maleT258 2.935247
                        0.330662 8.877
                                           <2e-16 ***
## maleT260 3.979770 0.311631 12.771
                                           <2e-16 ***
```

```
-0.127369 0.062492 -2.038
                                           0.0415 *
## I(x1^2) -0.005606 0.003320 -1.689
                                          0.0913 .
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Correlation matrix not shown by default, as p = 14 > 12.
## Use print(summary(glmer_nb), correlation=TRUE) or
      vcov(summary(glmer_nb))
                                      if you need it
## optimizer (bobyqa) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')
## unused control arguments ignored
## [1] "temp_ref = 35 glmer_nb re = quad"
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.0064789 (tol = 0.002, component 1)
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unide:
## - Rescale variables?
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.00834887 (tol = 0.002, component 1)
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unide:
## - Rescale variables?; Model is nearly unidentifiable: large eigenvalue ratio
## - Rescale variables?
## Warning in optTheta(g1, interval = interval, tol = tol, verbose = verbose, :
## unused control arguments ignored
## boundary (singular) fit: see help('isSingular')
## Generalized linear mixed model fit by maximum likelihood (Laplace
    Approximation) [glmerMod]
## Family: Negative Binomial(4.9599) ( log )
## Formula: song_count \sim -1 + male + x1 + I(x1^2) + (-1 + I(x1^2) || male)
     Data: data
##
```

```
## Control:
## glmerControl(boundary.tol = 0, tolPwrss = 0.1, optCtrl = list(maxiter = 1e+05,
       maxfun = 2e+06, trace = trace), optimizer = optimizer)
##
##
       ATC
                BIC
                      logLik deviance df.resid
##
      680.2
              714.0
                      -324.1
                                648.2
## Scaled residuals:
      Min
               1Q Median
                               30
                                      Max
## -1.9724 -0.6915 -0.1994 0.5434 3.1766
## Random effects:
## Groups Name
                  Variance Std.Dev.
## male
          I(x1^2) 0
## Number of obs: 61, groups: male, 12
##
## Fixed effects:
##
            Estimate Std. Error z value Pr(>|z|)
## maleT229 4.216007
                       0.297743 14.160
                                          <2e-16 ***
                       0.299006 14.190
## maleT231 4.242947
                                          <2e-16 ***
                      0.301998 11.586
## maleT234 3.499089
                                          <2e-16 ***
## maleT235 4.844355
                       0.313310 15.462
                                          <2e-16 ***
                       0.297332 14.740
## maleT236 4.382738
                                          <2e-16 ***
## maleT237 4.209983
                       0.306672 13.728
                                          <2e-16 ***
## maleT243 4.289398 0.285830 15.007
                                          <2e-16 ***
## maleT244 3.941226
                       0.321803 12.247
                                          <2e-16 ***
## maleT247 4.144315
                       0.301920 13.727
                                          <2e-16 ***
## maleT257 4.914289
                       0.286369 17.161
                                          <2e-16 ***
## maleT258 2.869099
                       0.327170 8.769
                                          <2e-16 ***
## maleT260 3.910894
                       0.308047 12.696
                                          <2e-16 ***
                       0.061788 -2.324
## x1
            -0.143609
                                          0.0201 *
## I(x1^2) -0.006424
                       0.003283 -1.957
                                          0.0504 .
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Correlation matrix not shown by default, as p = 14 > 12.
## Use print(summary(glmer_nb), correlation=TRUE) or
       vcov(summary(glmer_nb))
                                     if you need it
## optimizer (bobyqa) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')
## unused control arguments ignored
##
## [1] "temp ref = 35 glmer nb re = both"
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
```

```
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in optTheta(g1, interval = interval, tol = tol, verbose = verbose, :
## unused control arguments ignored
## boundary (singular) fit: see help('isSingular')
## Generalized linear mixed model fit by maximum likelihood (Laplace
    Approximation) [glmerMod]
## Family: Negative Binomial(5.2598) ( log )
## Formula: song_count \sim -1 + male + x1 + I(x1^2) + (-1 + x1 + I(x1^2) ||
##
      male)
     Data: data
##
## Control:
  glmerControl(boundary.tol = 0, tolPwrss = 0.1, optCtrl = list(maxiter = 1e+05,
##
      maxfun = 2e+06, trace = trace), optimizer = optimizer)
##
##
                BIC
                      logLik deviance df.resid
       AIC
##
     696.2
              732.1
                      -331.1
                                662.2
##
## Scaled residuals:
      Min
               1Q Median
                                     Max
## -2.0414 -0.7625 -0.2002 0.4121
                                 2.2875
## Random effects:
## Groups Name
                  Variance Std.Dev.
## male
          x1
                  1.045e-03 3.232e-02
## male.1 I(x1^2) 2.891e-11 5.377e-06
## Number of obs: 61, groups: male, 12
##
## Fixed effects:
             Estimate Std. Error z value Pr(>|z|)
## maleT229 4.4518665 0.2984930 14.914 < 2e-16 ***
## maleT231 4.1955448 0.2744380 15.288
                                         < 2e-16 ***
## maleT234 3.8829865 0.2980811 13.027
                                         < 2e-16 ***
## maleT235 4.7403106 0.2999752 15.802
                                         < 2e-16 ***
## maleT236 4.0008904 0.3171432 12.615
                                         < 2e-16 ***
## maleT237 4.8246073 0.3442351 14.015
                                         < 2e-16 ***
## maleT243 4.7482484
                      0.2767484
                                17.157
                                         < 2e-16 ***
## maleT244 3.8724934
                      0.3140062 12.333
                                         < 2e-16 ***
## maleT247 4.5707888 0.3018051 15.145
                                         < 2e-16 ***
## maleT257 5.3796854 0.2805020 19.179
                                         < 2e-16 ***
## maleT258 2.9799895
                      0.3276061
                                  9.096
                                         < 2e-16 ***
## maleT260 4.3916282 0.2994277 14.667
                                        < 2e-16 ***
           ## I(x1^2) -0.0044420 0.0001864 -23.834 < 2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Correlation matrix not shown by default, as p = 14 > 12.
## Use print(summary(glmer nb), correlation=TRUE) or
      vcov(summary(glmer_nb))
##
                                    if you need it
```

```
## optimizer (bobyqa) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')
## unused control arguments ignored
```

- glmer_poisson and glmer_nb only converge under certain re and only when when optimizer = "bobyqa".
- glmer_poisson summary

```
    re = "linear": Singular/boundary fit
    re = "quad": Fails to converge
    re = "both": Converges, overdisp.glmer indicates data is greatly overdispersed.
```

• glmer_nb summary

```
- re = "linear": Singular/boundary fit
- re = "quad": Singular/boundary fit
- re = "both": Singular/boundary fit
```

- It's note worthy that glmer_nb gives a singular value for the variance (i.e. effectively 0) for x1 using optimizer = "bobyqa", but gives the variance (i.e. effectively 0) for x1^2 using optimizer = "nloptwrap". This seems consistent with them being highly correlated as suggested in the || fittings.
- The x1^2 term is often (always?) non-significant in the nb models

| Formulation

```
est_cov <- TRUE
verbose <- 0
trace <- FALSE
optimizer <- "bobyqa" #"nloptwrap" # "bobyqa"
## Fit random effects models
for(dist in c("poisson", "nb")) {
    for(temp ref in c(35)) { # vec temp ref ){
        for(re in c("linear", "quad", "both")) {
            switch(re,
                    both = (formula_re <- song_count \sim -1 + male + x1 + I(x1^2) + (-1 + x1 + I(x1^2) |
                    linear = (formula_re <- song_count \sim -1 + male + x1 + I(x1^2) + (-1 + x1 | male)),
                    quad = (formula_re <- song_count \sim -1 + male + \times 1 + I(\times 1^2) + (-1 + I(\times 1^2) | male
            if(dist == "poisson"){
                print(paste("temp_ref =", temp_ref,"glmer_poisson", "re = ", re))
                glmer_poisson <- glmer(formula = formula_re,</pre>
                                        data = data,
                                         control = glmerControl(
                                             optCtrl = list(maxiter = 1E5,
                                                            maxfun = 2E6,
                                                            trace = trace),
                                             optimizer=optimizer),
                                         family = poisson(link = "log"),
```

```
verbose = verbose
                                       )
                print(summary(glmer_poisson))
                plot(glmer_poisson, ask=FALSE)
                fit_db[fit_db$temp_ref==temp_ref & fit_db$model == "glmer" & fit_db$dist == "poisson"
                print("evaluate overdispersion")
                glmer_qpoisson <- overdisp.glmer(glmer_poisson)</pre>
            }
            if(dist == "nb"){
                print(paste("temp_ref =", temp_ref,"glmer_nb", "re = ", re))
                try(glmer_nb <-
                        glmer.nb(formula = formula_re,
                                 data = data,
                                 ## control values are used by the initial optimization
                                 ## using a poisson glmer model, which doesn't converge
                                 control = glmerControl(
                                     boundary.tol = 0,
                                     tolPwrss=1e-1,
                                     optCtrl = list(maxiter = 1E5,
                                                    maxfun = 2E6,
                                                    trace = trace),
                                     optimizer = optimizer),
                                 ## nb.control values are used by the second optimizer
                                 ## Note need to set own optCtrl values
                                 nb.control = list(
                                     optCtrl = list(maxit = 1000,
                                                    maxfun = 2E5)),
                                 verbose = verbose
                                 )
                    )
                plot(glmer_nb, ask = FALSE)
                print(summary(glmer_nb))
                fit_db[fit_db$temp_ref==temp_ref & fit_db$model == "glmer" & fit_db$dist == "nb" & fi
           }
       }
   }
## [1] "temp_ref = 35 glmer_poisson re = linear"
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
```

```
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.015591 (tol = 0.002, component 1)
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unide:
## - Rescale variables?
## Generalized linear mixed model fit by maximum likelihood (Laplace
    Approximation) [glmerMod]
   Family: poisson (log)
## Formula: song_count ~ -1 + male + x1 + I(x1^2) + (-1 + x1 | male)
     Data: data
  Control: glmerControl(optCtrl = list(maxiter = 1e+05, maxfun = 2e+06,
##
      trace = trace), optimizer = optimizer)
##
##
       ATC
                BTC
                      logLik deviance df.resid
##
    1431.7
             1463.3
                      -700.8
                               1401.7
##
## Scaled residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
  -9.4099 -2.3601 -0.3221 1.7150 14.1120
##
## Random effects:
## Groups Name Variance Std.Dev.
              0.002995 0.05473
          x1
## Number of obs: 61, groups: male, 12
##
## Fixed effects:
            Estimate Std. Error z value Pr(>|z|)
## maleT229 3.958900 0.091330 43.347 < 2e-16 ***
## maleT231 4.240132 0.089705 47.267 < 2e-16 ***
## maleT234 3.758839 0.115413 32.569 < 2e-16 ***
## maleT235 4.504916 0.082918 54.330 < 2e-16 ***
## maleT236 3.876809
                       0.112270 34.531
                                        < 2e-16 ***
## maleT237 4.058857 0.101564 39.964 < 2e-16 ***
## maleT243 4.994830 0.081779 61.077 < 2e-16 ***
## maleT244 3.852043 0.121522 31.698 < 2e-16 ***
## maleT247 4.406083 0.089316 49.331
                                        < 2e-16 ***
## maleT257 5.526669 0.063079 87.615 < 2e-16 ***
## maleT258 2.833279
                       0.156104 18.150 < 2e-16 ***
                       0.113902 41.222 < 2e-16 ***
## maleT260 4.695259
                       0.021028 -5.616 1.95e-08 ***
## x1
           -0.118092
## I(x1^2) -0.005618
                       0.000760 -7.393 1.44e-13 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Correlation matrix not shown by default, as p = 14 > 12.
## Use print(summary(glmer_poisson), correlation=TRUE) or
      vcov(summary(glmer_poisson))
                                          if you need it
## optimizer (bobyqa) convergence code: 0 (OK)
## Model failed to converge with max|grad| = 0.015591 (tol = 0.002, component 1)
## Model is nearly unidentifiable: very large eigenvalue
```

```
## - Rescale variables?
## unused control arguments ignored
## [1] "evaluate overdispersion"
## Residual deviance: 939.335 on 46 degrees of freedom (ratio: 20.42)
## [1] "temp_ref = 35 glmer_poisson re = quad"
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.00649726 (tol = 0.002, component 1)
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unide:
## - Rescale variables?
## Generalized linear mixed model fit by maximum likelihood (Laplace
    Approximation) [glmerMod]
## Family: poisson (log)
## Formula: song_count \sim -1 + male + x1 + I(x1^2) + (-1 + I(x1^2) | male)
     Data: data
## Control: glmerControl(optCtrl = list(maxiter = 1e+05, maxfun = 2e+06,
##
      trace = trace), optimizer = optimizer)
##
##
       AIC
                BIC
                      logLik deviance df.resid
##
    1451.5
             1483.1
                      -710.7
                               1421.5
##
## Scaled residuals:
               1Q Median
                               3Q
      Min
                                      Max
## -8.9198 -2.4339 -0.2126 1.5026 14.6714
##
## Random effects:
## Groups Name
                  Variance Std.Dev.
## male I(x1^2) 8.531e-06 0.002921
## Number of obs: 61, groups: male, 12
##
## Fixed effects:
            Estimate Std. Error z value Pr(>|z|)
## maleT229 4.124778
                       0.074827 55.124 < 2e-16 ***
## maleT231 4.266849
                       0.074674 57.140 < 2e-16 ***
## maleT234 3.743742
                       0.090030 41.583 < 2e-16 ***
## maleT235 4.692110
                       0.069463 67.549 < 2e-16 ***
## maleT236 4.230641
                       0.083474 50.682 < 2e-16 ***
## maleT237 4.179025
                       0.080902 51.656 < 2e-16 ***
## maleT243 4.726213
                       0.067627 69.886 < 2e-16 ***
                       0.090476 43.952 < 2e-16 ***
## maleT244 3.976604
## maleT247 4.401292
                       0.073165 60.155
                                         < 2e-16 ***
                       0.056622 93.802 < 2e-16 ***
## maleT257 5.311227
## maleT258 3.084881
                       0.117646 26.222 < 2e-16 ***
## maleT260 4.344404 0.084856 51.197 < 2e-16 ***
```

```
-0.114833
                       0.014031 -8.184 2.74e-16 ***
## I(x1^2) -0.005631 0.001153 -4.884 1.04e-06 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Correlation matrix not shown by default, as p = 14 > 12.
## Use print(summary(glmer_poisson), correlation=TRUE) or
       vcov(summary(glmer_poisson))
                                          if you need it
## optimizer (bobyqa) convergence code: 0 (OK)
## Model failed to converge with max|grad| = 0.00649726 (tol = 0.002, component 1)
## Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
## unused control arguments ignored
## [1] "evaluate overdispersion"
## Residual deviance: 968.391 on 46 degrees of freedom (ratio: 21.052)
## [1] "temp_ref = 35 glmer_poisson re = both"
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.0254573 (tol = 0.002, component 1)
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unide:
## - Rescale variables?
## Generalized linear mixed model fit by maximum likelihood (Laplace
    Approximation) [glmerMod]
## Family: poisson (log)
## Formula: song_count ~ -1 + male + x1 + I(x1^2) + (-1 + x1 + I(x1^2) |
##
      male)
      Data: data
## Control: glmerControl(optCtrl = list(maxiter = 1e+05, maxfun = 2e+06,
      trace = trace), optimizer = optimizer)
##
##
        AIC
                BIC
                      logLik deviance df.resid
##
     1293.6 1329.5 -629.8
                              1259.6
##
## Scaled residuals:
      Min
              1Q Median
                               3Q
## -8.2553 -1.9791 -0.2244 1.2558 15.0288
##
## Random effects:
## Groups Name
                  Variance Std.Dev. Corr
## male x1
                  0.099106 0.31481
          I(x1<sup>2</sup>) 0.000276 0.01661 0.98
##
```

```
## Number of obs: 61, groups: male, 12
##
## Fixed effects:
            Estimate Std. Error z value Pr(>|z|)
## maleT229 4.157936 0.151855 27.381 <2e-16 ***
## maleT231 4.743229 0.147972 32.055
                                         <2e-16 ***
## maleT234 3.251286 0.221754 14.662 <2e-16 ***
## maleT235 5.107784 0.143705 35.544 <2e-16 ***
## maleT236 3.095568 0.189778 16.311
                                         <2e-16 ***
## maleT237 4.389249 0.180169 24.362
                                        <2e-16 ***
## maleT243 5.030252 0.149097 33.738
                                        <2e-16 ***
## maleT244 3.857537 0.293272 13.153
                                         <2e-16 ***
## maleT247 3.763094 0.189936 19.812
                                         <2e-16 ***
                                         <2e-16 ***
## maleT257 5.671076 0.112336 50.483
## maleT258 0.695339 0.424577
                                1.638
                                          0.101
## maleT260 6.912868
                       0.354188 19.518
                                         <2e-16 ***
           -0.111849
## x1
                       0.092585 -1.208
                                          0.227
## I(x1^2) -0.004894 0.004895 -1.000
                                          0.317
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Correlation matrix not shown by default, as p = 14 > 12.
## Use print(summary(glmer_poisson), correlation=TRUE) or
      vcov(summary(glmer_poisson))
                                         if you need it
## optimizer (bobyqa) convergence code: 0 (OK)
## Model failed to converge with max|grad| = 0.0254573 (tol = 0.002, component 1)
## Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
## unused control arguments ignored
## [1] "evaluate overdispersion"
## Residual deviance: 724.424 on 44 degrees of freedom (ratio: 16.464)
## [1] "temp_ref = 35 glmer_nb re = linear"
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.063488 (tol = 0.002, component 1)
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unide:
## - Rescale variables?
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
```

```
## Warning in optTheta(g1, interval = interval, tol = tol, verbose = verbose, :
## unused control arguments ignored
## boundary (singular) fit: see help('isSingular')
## Generalized linear mixed model fit by maximum likelihood (Laplace
     Approximation) [glmerMod]
  Family: Negative Binomial(4.895) (log)
## Formula: song_count \sim -1 + male + x1 + I(x1^2) + (-1 + x1 | male)
     Data: data
##
## Control:
   glmerControl(boundary.tol = 0, tolPwrss = 0.1, optCtrl = list(maxiter = 1e+05,
##
      maxfun = 2e+06, trace = trace), optimizer = optimizer)
##
##
       AIC
                BIC
                      logLik deviance df.resid
##
      680.2
              714.0
                      -324.1
                                648.2
##
## Scaled residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
## -1.9753 -0.6912 -0.1580 0.5586 3.0699
##
## Random effects:
## Groups Name Variance Std.Dev.
              3.432e-11 5.859e-06
          x1
## Number of obs: 61, groups: male, 12
##
## Fixed effects:
            Estimate Std. Error z value Pr(>|z|)
## maleT229 4.275715 0.301350 14.189
                                         <2e-16 ***
## maleT231 4.301052 0.302327 14.227
                                          <2e-16 ***
## maleT234 3.557692 0.305281 11.654
                                          <2e-16 ***
## maleT235 4.907960 0.316681 15.498
                                          <2e-16 ***
                       0.300701 14.739
## maleT236 4.432163
                                          <2e-16 ***
## maleT237 4.262585 0.309520 13.772
                                          <2e-16 ***
## maleT243 4.355303 0.289683 15.035
                                          <2e-16 ***
## maleT244 4.002712 0.324817 12.323
                                          <2e-16 ***
## maleT247 4.216963
                       0.305795 13.790
                                          <2e-16 ***
                                          <2e-16 ***
## maleT257 4.992306
                       0.291354 17.135
## maleT258 2.935247
                       0.330662
                                 8.877
                                          <2e-16 ***
## maleT260 3.979770
                       0.311631 12.771
                                          <2e-16 ***
                       0.062492 -2.038
## x1
            -0.127369
                                          0.0415 *
## I(x1<sup>2</sup>) -0.005606
                       0.003320 -1.689
                                          0.0913 .
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Correlation matrix not shown by default, as p = 14 > 12.
## Use print(summary(glmer_nb), correlation=TRUE) or
       vcov(summary(glmer_nb))
                                     if you need it
## optimizer (bobyqa) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')
## unused control arguments ignored
```

```
## [1] "temp_ref = 35 glmer_nb re = quad"
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.0064789 (tol = 0.002, component 1)
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unide:
## - Rescale variables?
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.00834887 (tol = 0.002, component 1)
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unide:
## - Rescale variables?; Model is nearly unidentifiable: large eigenvalue ratio
## - Rescale variables?
## Warning in optTheta(g1, interval = interval, tol = tol, verbose = verbose, :
## unused control arguments ignored
## boundary (singular) fit: see help('isSingular')
## Generalized linear mixed model fit by maximum likelihood (Laplace
    Approximation) [glmerMod]
## Family: Negative Binomial(4.9599) (log)
## Formula: song_count \sim -1 + male + x1 + I(x1^2) + (-1 + I(x1^2) | male)
##
     Data: data
## Control:
## glmerControl(boundary.tol = 0, tolPwrss = 0.1, optCtrl = list(maxiter = 1e+05,
##
       maxfun = 2e+06, trace = trace), optimizer = optimizer)
##
##
       AIC
                BIC
                       logLik deviance df.resid
##
      680.2
              714.0
                      -324.1
                                 648.2
##
## Scaled residuals:
##
      Min
                1Q Median
                                3Q
## -1.9724 -0.6915 -0.1994 0.5434 3.1766
##
## Random effects:
## Groups Name
                  Variance Std.Dev.
```

```
## male I(x1^2) 0
## Number of obs: 61, groups: male, 12
## Fixed effects:
            Estimate Std. Error z value Pr(>|z|)
## maleT229 4.216007 0.297743 14.160
                                          <2e-16 ***
## maleT231 4.242947
                       0.299006 14.190
                                          <2e-16 ***
## maleT234 3.499089 0.301998 11.586
                                          <2e-16 ***
## maleT235 4.844355
                       0.313310 15.462
                                          <2e-16 ***
## maleT236 4.382738 0.297332 14.740
                                         <2e-16 ***
## maleT237 4.209983
                      0.306672 13.728
                                         <2e-16 ***
## maleT243 4.289398 0.285830 15.007
                                          <2e-16 ***
## maleT244 3.941226 0.321803 12.247
                                         <2e-16 ***
## maleT247 4.144315 0.301920 13.727
                                         <2e-16 ***
## maleT257 4.914289
                       0.286369 17.161
                                          <2e-16 ***
## maleT258 2.869099
                       0.327170 8.769
                                          <2e-16 ***
                       0.308047 12.696
## maleT260 3.910894
                                          <2e-16 ***
        -0.143609
                       0.061788 -2.324
                                          0.0201 *
## I(x1^2) -0.006424
                       0.003283 -1.957
                                          0.0504 .
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Correlation matrix not shown by default, as p = 14 > 12.
## Use print(summary(glmer_nb), correlation=TRUE) or
      vcov(summary(glmer_nb))
##
                                     if you need it
## optimizer (bobyqa) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')
## unused control arguments ignored
## [1] "temp_ref = 35 glmer_nb re = both"
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.0262617 (tol = 0.002, component 1)
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unide:
## - Rescale variables?
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in (function (npt = min(n + 2L, 2L * n), rhobeg = NA, rhoend = NA, :
## unused control arguments ignored
## Warning in optTheta(g1, interval = interval, tol = tol, verbose = verbose, :
## unused control arguments ignored
```

```
## Warning in optTheta(g1, interval = interval, tol = tol, verbose = verbose, :
## Model failed to converge with max|grad| = 1.11108 (tol = 0.002, component 1)
## Warning in optTheta(g1, interval = interval, tol = tol, verbose = verbose, : Model is nearly unident
## - Rescale variables?; Model is nearly unidentifiable: large eigenvalue ratio
## - Rescale variables?
## Generalized linear mixed model fit by maximum likelihood (Laplace
    Approximation) [glmerMod]
   Family: Negative Binomial(25.2301) (log)
## Formula: song_count ~ -1 + male + x1 + I(x1^2) + (-1 + x1 + I(x1^2) |
##
      male)
##
     Data: data
## Control:
  glmerControl(boundary.tol = 0, tolPwrss = 0.1, optCtrl = list(maxiter = 1e+05,
      maxfun = 2e+06, trace = trace), optimizer = optimizer)
##
##
       AIC
                      logLik deviance df.resid
                      -353.8
##
     743.7
              781.7
                                707.7
##
## Scaled residuals:
               10 Median
      Min
                               3Q
                                      Max
## -3.0202 -0.8896 -0.2093 0.5991
                                  4.6854
## Random effects:
## Groups Name
                  Variance Std.Dev. Corr
                  0.0806149 0.28393
   male
          x1
          I(x1^2) 0.0002113 0.01454 0.98
## Number of obs: 61, groups: male, 12
## Fixed effects:
##
            Estimate Std. Error z value Pr(>|z|)
## maleT229 4.080355 0.324722 12.566 < 2e-16 ***
## maleT231 4.531626 0.384781 11.777 < 2e-16 ***
## maleT234 3.264457 0.388690
                                 8.399
                                        < 2e-16 ***
## maleT235 4.761868 0.462525 10.295 < 2e-16 ***
## maleT236 2.989138 0.500802
                                 5.969 2.39e-09 ***
## maleT237 4.288885 0.401619 10.679 < 2e-16 ***
## maleT243 5.012198 0.373985 13.402 < 2e-16 ***
## maleT244 3.725739 0.598097
                                  6.229 4.68e-10 ***
## maleT247 3.747123
                       0.432882
                                 8.656
                                        < 2e-16 ***
## maleT257 5.722484
                       0.356312 16.060
                                        < 2e-16 ***
## maleT258 0.671660
                       0.589491
                                 1.139
                                           0.255
                                 8.931
## maleT260 6.428693
                       0.719856
                                        < 2e-16 ***
## x1
           -0.143488
                       0.090652 -1.583
                                           0.113
## I(x1^2) -0.006533
                       0.004714 - 1.386
                                           0.166
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Correlation matrix not shown by default, as p = 14 > 12.
## Use print(summary(glmer_nb), correlation=TRUE) or
      vcov(summary(glmer_nb))
                                     if you need it
```

```
## optimizer (bobyqa) convergence code: 0 (OK)
## Model failed to converge with max|grad| = 1.11108 (tol = 0.002, component 1)
## Model is nearly unidentifiable: very large eigenvalue
## - Rescale variables?
## Model is nearly unidentifiable: large eigenvalue ratio
## - Rescale variables?
## unused control arguments ignored
```

- Changing temp_ref alters the linear term, as expected, but has little impact on the overall model fit.
- fixed effects quoisson and nb indicates data is highly over dispersed
- poisson and nb RE models do not converge
 - With RE for both x1 and x1^2 we get failure to converge and the correlation between these two RE is 1 (this means there's really, at best, one RE|male rather than two).
 - With single RE we get return boundary(singular) error because the var of teh RE is essentially zero. This indicates the model is too complex given the data

Basically, I don't think this avenue is fruitful, but I do need to understand how the covariance matrix is used in these models - What is the difference between the | and | | models? - I believe the | is a correlated random effects approach - Houseman test is used for comparing FE and RE - Does one of these assume the random effects themselves vary between trials for the same male as some references suggest, i.e. the RE for x1 for male z varies between replicates? Trials? - Seems like we could compare a lmList result with the glmer

Additional Fixed Effects fits

Fit Fixed Effects models

42: ## 43: } ## ^

- poisson model does give significant effects, but qpoisson indicates data is overdispersed. When overdisperion is taken into account, significance disappears.
 - This occurs even if you're only using /male on one of the x1 terms.

Dispersion parameter is 4547!

End

```
knitr::knit_exit()
```