R-code for "No evidence of sex allocation among three sympatric plovers in Madagascar that vary in mating system"

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In this document we provide all the necessary code for reproducing the analyses presented in our paper. To access the dataset and Rmarkdown file, please download this GitHub repository. Simply follow the link and click on *Download ZIP* on the right-hand side of the page. An explanation of the files in the repository can be found in the Readme file. Please don't hesitate to contact Luke at luke.eberhart[at]gmail.com if you have any questions.

The structure of the code we present here follows the analyses presented in the Results section of the paper.

Prerequisites:

- For running the complete code you need a files subfolder containing the raw data downloaded from data folder provided in the GitHub repository.
- The following packages are needed for analysis and can be easily installed from CRAN by uncommenting the install.packages functions:

```
# install.packages("dplyr")
# install.packages("lme4")
# install.packages("ggplot2")
# install.packages("RColorBrewer")
# install.packages("grid")
library(dplyr)
library(lme4)
library(ggplot2)
library(RColorBrewer)
library(grid)
```

Loading and wrangling data

To start, please load the following dataset into your R environment:

• plover chick data contains the raw chick data collected from the field. Each row is a single capture of a chick at its nest. Each chick has a unique identifier chickID and each brood has a unique identifier nestID. The geographic coordinates are also shown as easting and northing (Coordinate System: UTM Zone 38S Tananarive 1925) which are presented in the map to illustrate the spatial distribution of the nests used in this study. A chick can be one of three plover species (Kittlitz's, Madagascar, or White-fronted). year indicates the year during which an individual was captured and dateinseason indicates the number of days since the start of the breeding season that an individual hatched. The dummy variables male and female indicate the molecular sex-type of an individual with "1" signifing the sex.

Define the variables of the plover_chick correctly: easting and northing need to recognized as numeric, and chickID, nestID, species, and year need to recognized as factors.

```
plover_chick[,c("easting", "northing")] <-
    lapply(plover_chick[,c("easting", "northing")], as.numeric)

plover_chick[,c("chickID", "nestID", "species", "year")] <-
    lapply(plover_chick[,c("chickID", "nestID", "species", "year")], as.factor)</pre>
```

Summarize the plover_chick data such that each row represents data for a nest.

Do hatching sex ratios deviate from parity?

Linear regression mixed models for each species predicting deviation in hatching sex ratio while controlling for non-independence in siblings (i.e. random effect: nestID).

• Kittlitz's plover

```
KiP_HSR_model <- lme4::glmer(cbind(no_males, no_females) ~ (1| nestID),</pre>
                          data = dplyr::filter(plover_nest, species == "Kittlitz's"),
                          family = binomial)
summary(KiP_HSR_model)
#> Generalized linear mixed model fit by maximum likelihood (Laplace
#>
    Approximation) [qlmerMod]
#> Family: binomial ( logit )
#> Formula: cbind(no_males, no_females) ~ (1 | nestID)
      Data: dplyr::filter(plover_nest, species == "Kittlitz's")
#>
#>
#>
                BIC logLik deviance df.resid
       AIC
#>
      153.3
               157.8
                      -74.6 149.3
#>
```

• Madagascar plover

```
MP HSR model <- lme4::glmer(cbind(no males, no females) ~ (1 | nestID),
                         data = dplyr::filter(plover_nest, species == "Madagascar"),
                         family = binomial)
summary(MP_HSR_model)
#> Generalized linear mixed model fit by maximum likelihood (Laplace
   Approximation) [qlmerMod]
#> Family: binomial ( logit )
#> Formula: cbind(no_males, no_females) ~ (1 | nestID)
     Data: dplyr::filter(plover_nest, species == "Madagascar")
#>
#>
       AIC
                BIC logLik deviance df.resid
      21.8
               22.5
                       -8.9
                               17.8
#>
#>
#> Scaled residuals:
      Min 1Q Median
                             3Q
#> -0.4009 -0.4009 -0.4009 0.3341 1.0690
#>
#> Random effects:
#> Groups Name
                      Variance Std. Dev.
#> nestID (Intercept) 0
                             0
#> Number of obs: 11, groups: nestID, 11
#> Fixed effects:
              Estimate Std. Error z value Pr(>|z|)
#> (Intercept) 0.5596 0.4432 1.263 0.207
```

White-fronted plover

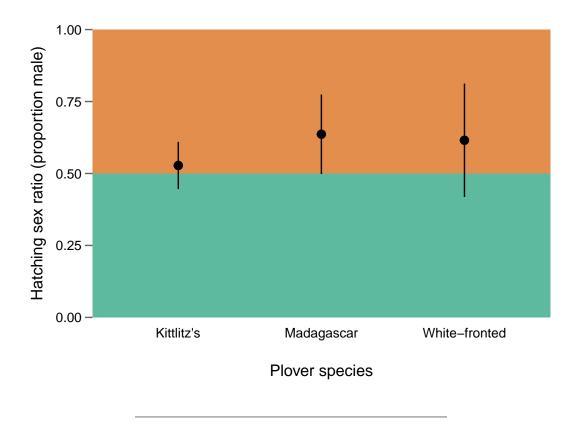
```
#>
#>
       AIC
               BIC logLik deviance df.resid
               35.4
      34.3
                      -15.1
#>
                                30.3
#>
#> Scaled residuals:
      Min
#>
             1Q Median
                             3Q
#> -1.1406 -0.2352 -0.2352 0.7478 0.8308
#>
#> Random effects:
#> Groups Name
                      Variance Std. Dev.
                             1.253
#> nestID (Intercept) 1.57
#> Number of obs: 13, groups: nestID, 13
#>
#> Fixed effects:
#>
              Estimate Std. Error z value Pr(>|z|)
#> (Intercept) 0.5887 0.5923 0.994
```

Summarize the plover_nest data such that each row represents a species.

Plot the species summary.

```
# draw the blank plot for the background
Figure 1 blank <-
  ggplot2::ggplot(data = plover_summary, aes(y = mean, x = species)) +
            theme bw() +
            annotate("rect", xmin = 0, xmax = 6, ymin = 0, ymax = 0.5, alpha = 0.7,
                     fill = RColorBrewer::brewer.pal(8, "Dark2")[c(1)]) +
            annotate("rect", xmin = 0, xmax = 6, ymin = 0.5, ymax = 1, alpha = 0.7,
                     fill = RColorBrewer::brewer.pal(8, "Dark2")[c(2)]) +
                theme(text = element_text(colour = "white"),
                      legend.position = "none",
                      legend.background = element_rect(fill = NA),
                      axis.title.y = element_text(size = 12, margin = margin(0, 10, 0, 0)),
                      axis.text.y = element_text(size = 10),
                      axis.title.x = element_text(size = 12, margin = margin(20, 0, 0, 0)),
                      axis.text.x = element_text(size = 10),
                      axis.ticks.x = element_blank(),
                      axis.ticks.y = element_line(size = 0.5, colour = "white"),
                      axis.ticks.length = unit(0.2, "cm"),
                      panel.grid.major = element blank(),
                      panel.grid.minor = element_blank(),
```

```
panel.border = element_blank(),
                      panel.margin = unit(0.75, "lines"),
                      plot.margin = unit(c(0.5, 1.5, 0.5, 0), "cm")) +
                scale_y_continuous(limits = c(0, 1), expand = c(0, 0)) +
                scale_x_{continuous}(limits=c(0, 6), breaks=c(0, 1, 2), expand = c(0, 0)) +
                xlab("Plover species") +
                ylab("Hatching sex ratio (proportion male)")
# draw the plot with the data
Figure 1 <-
  ggplot2::ggplot(data = plover_summary, aes(y = mean, x = species)) +
            theme bw() +
            geom_pointrange(aes(ymin = lower_CI, ymax = upper_CI)) +
            theme(legend.position = "none",
                  panel.background = element_rect(fill = "transparent", colour = NA),
                  plot.background = element_rect(fill = "transparent",colour = NA),
                  axis.title.y = element_text(size = 12, margin = margin(0, 10, 0, 0)),
                  axis.text.y = element_text(size = 10),
                  axis.title.x = element_text(size = 12, margin = margin(20, 0, 0, 0)),
                  axis.text.x = element_text(size = 10),
                  axis.ticks.x = element_blank(),
                  axis.ticks.y = element_line(size = 0.5, colour = "grey40"),
                  axis.ticks.length = unit(0.2, "cm"),
                  panel.grid.major = element_blank(),
                  panel.grid.minor = element_blank(),
                  panel.border = element_blank(),
                  panel.margin = unit(0.75, "lines"),
                  plot.margin = unit(c(0.5, 1.5, 0.5, 0), "cm")) +
            scale_y_continuous(limits = c(0, 1), expand = c(0, 0)) +
            xlab("Plover species") +
            ylab("Hatching sex ratio (proportion male)")
# Print the two layers
grid::pushViewport(
  grid::viewport(
    layout = grid::grid.layout(1, 1, widths = unit(1, "npc"))))
print(Figure_1_blank, newpage = FALSE)
print(Figure_1, newpage = FALSE)
grid::popViewport()
```



Does hatching sex ratio vary seasonaly or annually?

Linear regression mixed models for each species predicting annual and seasonal deviations in hatching sex ratio while controlling for non-independence in siblings (i.e. random effect: nestID).

• Kittlitz's plover

```
# Annual model
KiP HSR year <-
  lme4::glmer(male ~ (1| nestID) + year,
              data = dplyr::filter(plover_chick, species == "Kittlitz's"),
              family = binomial)
# Seasonal model
KiP_HSR_season <-
  lme4::glmer(male ~ (1| nestID) + dateinseason,
              data = dplyr::filter(plover_chick, species == "Kittlitz's"),
              family = binomial)
# No temporal effect model
KiP_HSR_null <-</pre>
  lme4::glmer(male ~ (1| nestID),
              data = dplyr::filter(plover_chick, species == "Kittlitz's"),
              family = binomial)
# Assess the effect of year
anova(KiP_HSR_year, KiP_HSR_null)
```

```
#> Data: dplyr::filter(plover_chick, species == "Kittlitz's")
#> Models:
#> KiP_HSR_null: male ~ (1 | nestID)
#> KiP_HSR_year: male ~ (1 | nestID) + year
#>
              Df AIC BIC logLik deviance Chisq Chi Df Pr(>Chisq)
#> KiP_HSR_null 2 203.18 209.12 -99.591 199.18
#> KiP_HSR_year 5 205.68 220.53 -97.840 195.68 3.5016
                                                           3
                                                                 0.3206
# Assess the effect of season
anova(KiP_HSR_season, KiP_HSR_null)
#> Data: dplyr::filter(plover_chick, species == "Kittlitz's")
#> Models:
#> KiP_HSR_null: male ~ (1 | nestID)
#> KiP_HSR_season: male ~ (1 | nestID) + dateinseason
                Df AIC BIC logLik deviance Chisq Chi Df Pr(>Chisq)
#> KiP_HSR_null 2 203.18 209.12 -99.591 199.18
#> KiP_HSR_season 3 204.62 213.53 -99.312 198.62 0.5587
```

• Madagascar plover

```
# Annual model
MP_HSR_year <-
  lme4::glmer(male ~ (1| nestID) + year,
              data = dplyr::filter(plover_chick, species == "Madagascar"),
              family = binomial)
# Seasonal model
MP_HSR_season <-
  lme4::glmer(male ~ (1| nestID) + dateinseason,
              data = dplyr::filter(plover_chick, species == "Madagascar"),
              family = binomial)
# No temporal effect model
MP HSR null <-
  lme4::glmer(male ~ (1| nestID),
              data = dplyr::filter(plover chick, species == "Madagascar"),
              family = binomial)
# Assess the effect of year
anova(MP_HSR_year, MP_HSR_null)
#> Data: dplyr::filter(plover_chick, species == "Madagascar")
#> Models:
#> MP_HSR_null: male ~ (1 | nestID)
#> MP_HSR_year: male ~ (1 | nestID) + year
             Df AIC BIC logLik deviance Chisq Chi Df Pr(>Chisq)
#>
#> MP_HSR_null 2 32.841 35.023 -14.421 28.841
#> MP_HSR_year 5 37.253 42.708 -13.626 27.253 1.5886
                                                            3
                                                                    0.662
# Assess the effect of season
anova(MP_HSR_season, MP_HSR_null)
#> Data: dplyr::filter(plover_chick, species == "Madagascar")
#> Models:
#> MP_HSR_null: male ~ (1 | nestID)
```

```
#> MP_HSR_season: male ~ (1 | nestID) + dateinseason

#> Df AIC BIC logLik deviance Chisq Chi Df Pr(>Chisq)

#> MP_HSR_null 2 32.841 35.023 -14.421 28.841

#> MP_HSR_season 3 34.468 37.741 -14.234 28.468 0.3735 1 0.5411
```

• Madagascar plover

```
# Annual model
WfP_HSR_year <-
 lme4::glmer(male ~ (1| nestID) + year,
             data = dplyr::filter(plover_chick, species == "White-fronted"),
             family = binomial)
# Seasonal model
WfP_HSR_season <-
 lme4::glmer(male ~ (1| nestID) + dateinseason,
             data = dplyr::filter(plover_chick, species == "White-fronted"),
             family = binomial)
# No temporal effect model
WfP HSR null <-
 lme4::glmer(male ~ (1| nestID),
             data = dplyr::filter(plover_chick, species == "White-fronted"),
             family = binomial)
# Assess the effect of year
anova(WfP_HSR_year, WfP_HSR_null)
#> Data: dplyr::filter(plover_chick, species == "White-fronted")
#> Models:
#> WfP_HSR_null: male ~ (1 | nestID)
#> WfP_HSR_year: male ~ (1 | nestID) + year
              Df AIC
                          BIC logLik deviance Chisq Chi Df Pr(>Chisq)
#> WfP_HSR_null 2 42.606 45.408 -19.303 38.606
#> WfP_HSR_year 4 43.114 48.718 -17.557 35.114 3.4922
                                                          2
                                                                 0.1745
# Assess the effect of season
anova(WfP_HSR_season, WfP_HSR_null)
#> Data: dplyr::filter(plover_chick, species == "White-fronted")
#> Models:
#> WfP_HSR_null: male ~ (1 | nestID)
#> WfP_HSR_season: male ~ (1 | nestID) + dateinseason
                Df AIC BIC logLik deviance Chisq Chi Df Pr(>Chisq)
#> WfP_HSR_null 2 42.606 45.408 -19.303 38.606
#> WfP_HSR_season 3 44.107 48.310 -19.053 38.107 0.4992
```