## 10. Applying random forest on field data - gene

Fay

2022-11-04

#### Aim:

- Applying the models established in the script: 9
- How are hybrid mice different to the parental species?

## Load necessary libraries:

```
#install.packages("optima", version = "2021-10.12") # this package is required for
#the parasite load package to work
library(tidyverse)
library(tidyr)
library(dplyr)
library(cowplot)
library(randomForest)
library(ggplot2)
library(VIM) # visualizing missing data
library(mice) # imputing missing data without predictors
library(ggpubr)
library(optimx)
library(rfUtilities) # Implements a permutation test cross-validation for
# Random Forests models
library(mice) #imputations
library(fitdistrplus) #testing distributions
library(logspline)
library(caret)
```

#### Field data

#### Import field data

```
hm <- read.csv("output_data/imputed_mice.csv")</pre>
```

#### Clean data

```
Field <- hm %>%
  filter(origin == "Field") %>%
   drop_na(HI)
```

We have 1921 mice in total.

Prepare vectors for selecting

#### **Actual Cleaning**

```
#remove the unimputed columns
Field <- unique(Field) %>%
 dplyr::select(-all_of(Genes_wild)) %>%
 rename_with(~str_remove(., '_imp'))
#select the imputed gene columns
gene <- Field %>%
 dplyr::select(c(Mouse_ID, all_of(Genes_wild)))
genes <- gene %>%
 dplyr::select(-Mouse_ID)
#remove rows with only nas
genes <- genes[,colSums(is.na(genes))<nrow(genes)]</pre>
#remove colums with only nas
genes <- genes[rowSums(is.na(genes)) != ncol(genes), ]</pre>
# select the same rows from the gene data
gene <- gene[row.names(genes),]</pre>
# select the same rows from the field data
Field <- Field[row.names(genes),]</pre>
```

## Predicting weight loss in our imputed field data

Start by making the predictions for the field data.

```
# load predicting weight loss model
weight_loss_predict <- readRDS("r_scripts/models/predict_WL.rds")
set.seed(540)

#The predict() function in R is used to predict the values based on the input data.
predictions_field <- predict(weight_loss_predict, genes)

# assign test.data to a new object, so that we can make changes
result_field <- genes</pre>
```

```
#add the new variable of predictions to the result object
result_field <- cbind(result_field, predictions_field)

# add it to the field data
Field <- cbind(Field, predictions_field)</pre>
```

# It is time to apply the package of Alice Balard et al. on our predictions!

Let's see if we indeed have differences across the hybrid index with our predicted weight loss.

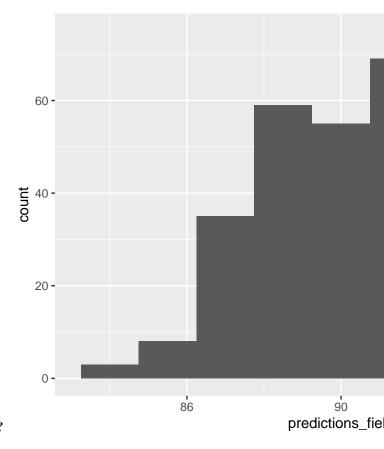
#### Install the package

```
## ggplot2 (5aee4a6ba... -> b27acf49c...) [GitHub]
##
# checking for file '/tmp/Rtmp4imQjc/remotesaf3aa460e5868/cran-ggplot2-b27acf4/DESCRIPTION' ... OK
## * preparing 'ggplot2':
## * checking DESCRIPTION meta-information ... OK
## * checking vignette meta-information ... OK
## * checking for LF line-endings in source and make files and shell scripts
## * checking for empty or unneeded directories
## * building 'ggplot2_3.4.0.tar.gz'
##
## * checking for file '/tmp/Rtmp4imQjc/remotesaf3aa210d18d/alicebalard-parasiteLoad-1b43216/DESCRIPTIO
## * preparing 'parasiteLoad':
## * checking DESCRIPTION meta-information ... OK
## * checking for LF line-endings in source and make files and shell scripts
## * checking for empty or unneeded directories
## * checking for empty or unneeded directories
## * building 'parasiteLoad_0.1.0.tar.gz'
```

#### Data diagnostics

#### Visualizations

```
Field %>% ggplot(aes(x = predictions_field)) +
  geom_histogram(binwidth = 1.5)
```



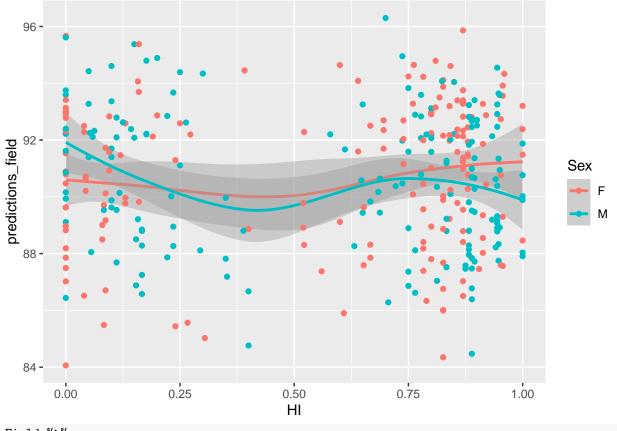
What is the distribution of the predicted weight loss?

Rough graph of our predictions against the hybrid index and against the

```
Field %>%
    ggplot(aes(x = HI , y = predictions_field , color = Sex)) +
    geom_smooth() +
    geom_point()
```

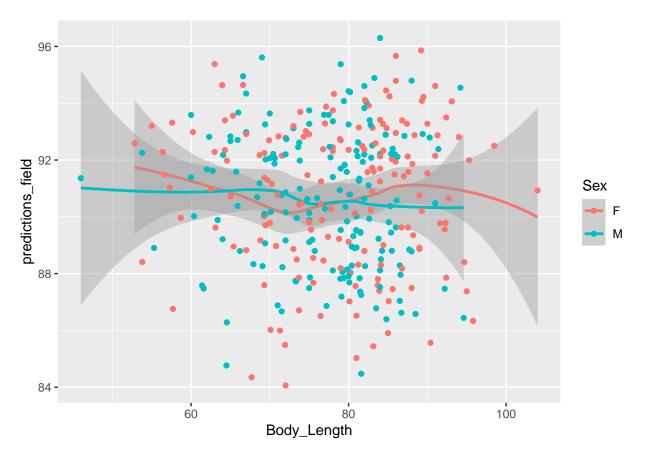
#### body length

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



```
Field %>%
    ggplot(aes(x = Body_Length , y = predictions_field , color = Sex)) +
    geom_smooth() +
    geom_point()
```

- ##  $geom_smooth()$  using method = 'loess' and formula 'y ~ x'
- ## Warning: Removed 1 rows containing non-finite values (stat\_smooth).
- ## Warning: Removed 1 rows containing missing values (geom\_point).



#### Fitting distributions??

Ratios / Percentages are not normally distributed. Weibull is a good distributions.

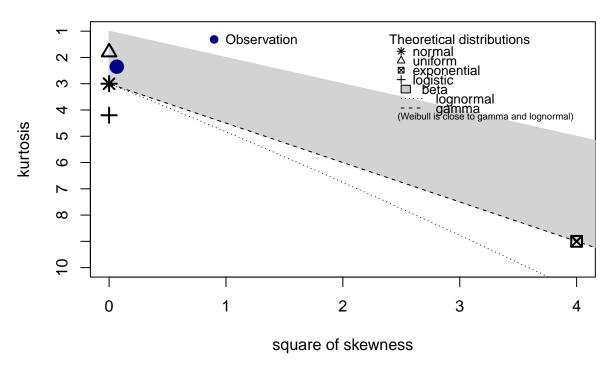
Alice used weibull for the qpcr data. (paper)

```
Field <- Field %>%
dplyr::mutate(WL = predictions_field)

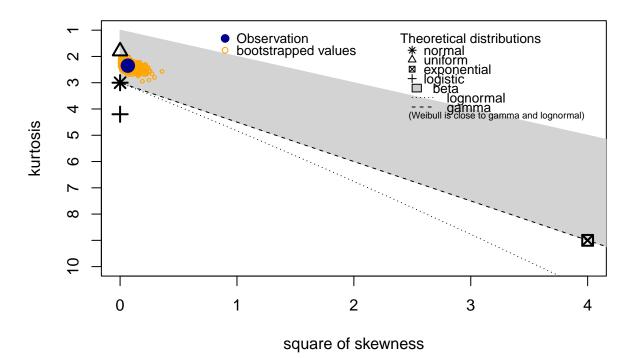
x <- Field$WL

descdist(data = x, discrete = FALSE)</pre>
```

# **Cullen and Frey graph**



## **Cullen and Frey graph**



```
## summary statistics
## -----
## min: 84.06162 max: 96.29867
## median: 90.93037
## mean: 90.69378
## estimated sd: 2.492399
## estimated skewness: -0.2572451
## estimated kurtosis: 2.350916
```

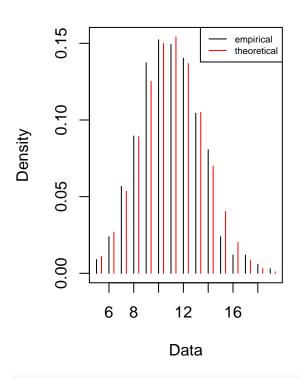
#### Test for binomial distribution

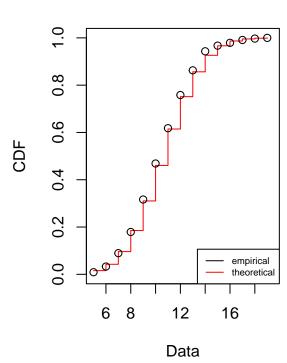
```
set.seed(10)
n = 25
size = 27
prob = .4
data = rbinom(x, size = size, prob = prob)
fit = fitdist(data = data, dist="binom",
                   fix.arg=list(size = size),
                   start=list(prob = 0.1))
summary(fit)
## Fitting of the distribution ' binom ' by maximum likelihood
## Parameters :
##
        estimate Std. Error
## prob 0.399558 0.005150141
## Fixed parameters:
        value
## size
           27
```

```
## Loglikelihood: -779.317 AIC: 1560.634 BIC: 1564.448 plot(fit)
```

## Emp. and theo. distr.

## Emp. and theo. CDFs





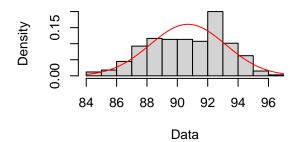
```
normal_ <- fitdist(x, "norm")</pre>
weibull <- fitdist(x, "weibull")</pre>
gamma_ <- fitdist(x, "gamma")</pre>
# Define function to be used to test, get the log lik and aic
tryDistrib <- function(x, distrib){</pre>
  # deals with fitdistr error:
  fit <- tryCatch(MASS::fitdistr(x, distrib), error=function(err) "fit failed")</pre>
  return(list(fit = fit,
               loglik = tryCatch(fit$loglik, error=function(err) "no loglik computed"),
               AIC = tryCatch(fit$aic, error=function(err) "no aic computed")))
}
findGoodDist <- function(x, distribs, distribs2){</pre>
  1 =lapply(distribs, function(i) tryDistrib(x, i))
  names(1) <- distribs</pre>
  print(1)
  listDistr <- lapply(distribs2, function(i){</pre>
    if (i %in% "t"){
      fitdistrplus::fitdist(x, i, start = list(df =2))
    } else {
      fitdistrplus::fitdist(x,i)
```

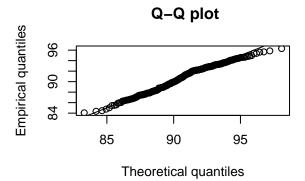
```
}}
  )
  par(mfrow=c(2,2))
  denscomp(listDistr, legendtext=distribs2)
  cdfcomp(listDistr, legendtext=distribs2)
  qqcomp(listDistr, legendtext=distribs2)
 ppcomp(listDistr, legendtext=distribs2)
 par(mfrow=c(1,1))
tryDistrib(x, "normal")
Functions for testing distributions
## $fit
##
         mean
                        sd
##
    90.69377795
                    2.48867665
## ( 0.13597093) ( 0.09614597)
##
## $loglik
## [1] -780.781
##
## $AIC
## NULL
tryDistrib(x, "binomial")
## $fit
## [1] "fit failed"
##
## $loglik
## [1] "no loglik computed"
## $AIC
## [1] "no aic computed"
tryDistrib(x, "student")
## $fit
## [1] "fit failed"
##
## $loglik
## [1] "no loglik computed"
##
## $AIC
## [1] "no aic computed"
tryDistrib(x, "weibull")
## $fit
##
       shape
                     scale
##
    41.4653052 91.8758382
## ( 1.7368487) ( 0.1279802)
##
## $loglik
## [1] -782.1914
```

```
##
## $AIC
## NULL
tryDistrib(x, "weibullshifted")
## $fit
## [1] "fit failed"
##
## $loglik
   [1] "no loglik computed"
##
##
## $AIC
## [1] "no aic computed"
findGoodDist(x, "normal", "weibull")
## $normal
   $normal$fit
##
##
          mean
                            sd
##
     90.69377795
                       2.48867665
    (0.13597093) (0.09614597)
##
##
## $normal$loglik
   [1] -780.781
##
## $normal$AIC
## NULL
     Histogram and theoretical densities
                                                            Empirical and theoretical CDFs
     0.15
Density
                                      weibull
                                                         9.0
                                                   CDF
     0.00
                                                                                          weibull
                                                         0.0
               86
                                                              84
                                                                   86
                                                                        88
          84
                    88
                          90
                               92
                                     94
                                          96
                                                                              90
                                                                                   92
                                                                                        94
                                                                                              96
                         data
                                                                             data
                      Q-Q plot
                                                                          P-P plot
                                                   Empirical probabilities
Empirical quantiles
     96
                                                        9.0
     90
                                   weibull
                                                                                       weibull
                                                         0.0
     84
             80
                      85
                               90
                                                                   0.2
                                                                          0.4
                                         95
                                                             0.0
                                                                                 0.6
                                                                                        8.0
                                                                                              1.0
                 Theoretical quantiles
                                                                    Theoretical probabilities
```

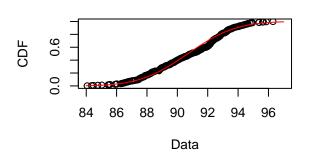
plot(normal\_)

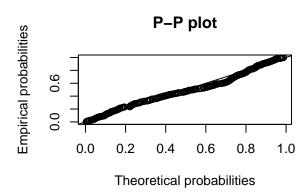
## Empirical and theoretical dens.





## **Empirical and theoretical CDFs**

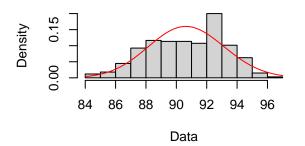


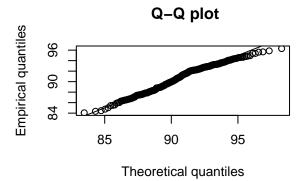


#### summary(normal\_)

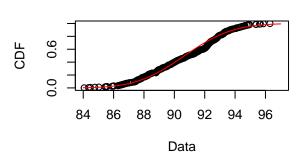
```
## Fitting of the distribution ' norm ' by maximum likelihood
## Parameters :
         estimate Std. Error
##
## mean 90.693778 0.1359709
## sd
         2.488677 0.0961459
## Loglikelihood: -780.781
                              AIC: 1565.562
                                               BIC: 1573.19
## Correlation matrix:
##
        mean sd
## mean
           1 0
## sd
           0
             1
plot(gamma_)
```

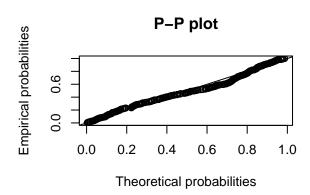
## Empirical and theoretical dens.





## **Empirical and theoretical CDFs**



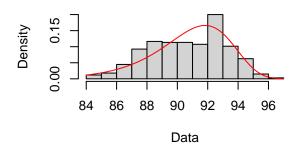


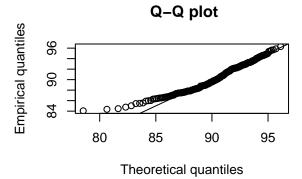
### summary(gamma\_)

```
\mbox{\tt \#\#} Fitting of the distribution ' gamma ' by maximum likelihood
## Parameters :
           estimate Std. Error
##
## shape 1321.02729 99.375775
  rate
           14.56579
                       1.095947
## Loglikelihood: -781.6076
                                AIC: 1567.215
                                                  BIC: 1574.843
## Correlation matrix:
##
             shape
## shape 1.0000000 0.9998004
## rate 0.9998004 1.0000000
```

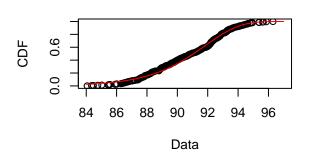
plot(weibull\_)

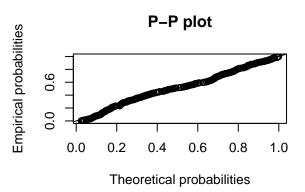
## Empirical and theoretical dens.





## **Empirical and theoretical CDFs**





summary(weibull\_)

```
## Fitting of the distribution ' weibull ' by maximum likelihood
## Parameters :
         estimate Std. Error
## shape 41.46612
                    1.736871
## scale 91.87592
                    0.127978
## Loglikelihood: -782.1914
                               AIC: 1568.383
                                                BIC: 1576.011
## Correlation matrix:
##
            shape
                     scale
## shape 1.000000 0.324389
## scale 0.324389 1.000000
```

#### Is alpha significant for each hypothesis?

We have a weibull distribution!

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

Field <- Field %>%
  mutate(WL = 100 - WL)

parasiteLoad::getParamBounds("weibull", data = Field, response = "WL")
```

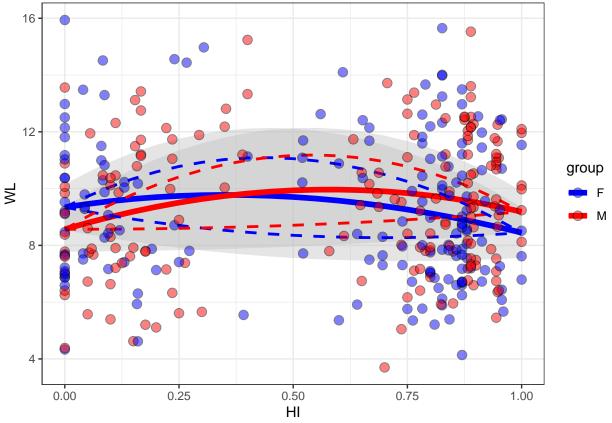
```
##
        L1start
                         L1LB
                                       L1UB
                                                  L2start
                                                                   L2LB
                                                                                 L2UB
    9.306222052 0.000000001 15.938381859
                                              9.306222052 0.000000001 15.938381859
##
                      alphaLB
                                    alphaUB myshapeStart
                                                              myshapeLB
                                                                            myshapeUB
     alphaStart
    0.000000000 -5.000000000 \ \ 5.000000000 \ \ 1.000000000 \ \ \ 0.000000001 \ \ 5.000000000
```

```
speparam \leftarrow c(L1start = 10,
                     L1LB = 1e-9,
                     L1UB = 20,
                     L2start = 10.
                     L2LB = 1e-9,
                     L2UB = 20,
                     alphaStart = 0, alphaLB = -5, alphaUB = 5,
                     myshapeStart = 1, myshapeLB = 1e-9, myshapeUB = 5)
##A11
fitWL_Sex <- parasiteLoad::analyse(data = Field,</pre>
                        response = "WL",
                        model = "weibull",
                        group = "Sex")
## [1] "Analysing data for response: WL"
## [1] "Fit for the response: WL"
## [1] "Fitting for all"
## [1] "Fitting model basic without alpha"
## [1] "Did converge"
## [1] "Fitting model basic with alpha"
## [1] "Did converge"
## [1] "Fitting model advanced without alpha"
## [1] "Did converge"
## [1] "Fitting model advanced with alpha"
## [1] "Did converge"
## [1] "Fitting for groupA : F"
## [1] "Fitting model basic without alpha"
## [1] "Did converge"
## [1] "Fitting model basic with alpha"
## [1] "Did converge"
## [1] "Fitting model advanced without alpha"
## [1] "Did converge"
## [1] "Fitting model advanced with alpha"
## [1] "Did converge"
## [1] "Fitting for groupB : M"
## [1] "Fitting model basic without alpha"
## [1] "Did converge"
## [1] "Fitting model basic with alpha"
## [1] "Did converge"
## [1] "Fitting model advanced without alpha"
## [1] "Did converge"
## [1] "Fitting model advanced with alpha"
## [1] "Did converge"
## [1] "Testing HO no alpha vs alpha"
      dLL dDF
                 pvalue
            1 0.1283225
## 1 1.16
## [1] "Testing H1 no alpha vs alpha"
##
     dLL dDF
                pvalue
## 1 1.23
            1 0.116121
## [1] "Testing H2 groupA no alpha vs alpha"
   dLL dDF
                pvalue
## 1 0.2
           1 0.5246522
## [1] "Testing H2 groupB no alpha vs alpha"
```

```
##
      dLL dDF
                 pvalue
            1 0.1011852
## 1 1.34
  [1] "Testing H3 groupA no alpha vs alpha"
      dLL dDF
                 pvalue
##
## 1 0.75
            1 0.2210683
  [1] "Testing H3 groupB no alpha vs alpha"
      dLL dDF
                  pvalue
## 1 1.56
            1 0.07695409
## [1] "Testing H1 vs H0"
                pvalue
     dLL dDF
## 1 0.1
           1 0.6624668
## [1] "Testing H2 vs H0"
                 pvalue
      dLL dDF
            3 0.4306275
## 1 1.38
## [1] "Testing H3 vs H1"
##
      dLL dDF
                 pvalue
## 1 3.72
            4 0.1140242
## [1] "Testing H3 vs H2"
      dLL dDF
                  pvalue
## 1 2.44
            2 0.08709369
fitWL Sex
## $HO
##
## Call:
## bbmle::mle2(minuslogl = response ~ dweibull(shape = myshape,
##
       scale = MeanLoad(L1, L1, alpha, HI)/gamma(1 + (1/myshape))),
       start = start, method = config$method, optimizer = config$optimizer,
##
##
       data = data, lower = c(L1 = paramBounds[["L1LB"]], alpha = paramBounds[["alphaLB"]],
##
           myshape = paramBounds[["myshapeLB"]]), upper = c(L1 = paramBounds[["L1UB"]],
##
           alpha = paramBounds[["alphaUB"]], myshape = paramBounds[["myshapeUB"]]),
##
       control = config$control)
##
##
   Coefficients:
##
           T.1
                   alpha
                             myshape
    9.0079495 -0.1450148
                          4.1118351
##
## Log-likelihood: -780.86
## Best method: L-BFGS-B
##
## $H1
##
## Call:
## bbmle::mle2(minuslog1 = response ~ dweibull(shape = myshape,
##
       scale = MeanLoad(L1, L2, alpha, HI)/gamma(1 + (1/myshape))),
##
       start = start, method = config$method, optimizer = config$optimizer,
       data = data, lower = c(L1 = paramBounds[["L1LB"]], L2 = paramBounds[["L2LB"]],
##
##
           alpha = paramBounds[["alphaLB"]], myshape = paramBounds[["myshapeLB"]]),
       upper = c(L1 = paramBounds[["L1UB"]], L2 = paramBounds[["L2UB"]],
##
##
           alpha = paramBounds[["alphaUB"]], myshape = paramBounds[["myshapeUB"]]),
##
       control = config$control)
##
  Coefficients:
                      L2
##
           L1
                               alpha
                                        myshape
```

```
9.0783810 8.9308463 -0.1526132 4.1153069
##
## Log-likelihood: -780.77
## Best method: bobyqa
## $H2
## $H2$groupA
##
## Call:
  bbmle::mle2(minuslog1 = response ~ dweibull(shape = myshape,
##
       scale = MeanLoad(L1, L1, alpha, HI)/gamma(1 + (1/myshape))),
       start = start, method = config$method, optimizer = config$optimizer,
##
       data = data, lower = c(L1 = paramBounds[["L1LB"]], alpha = paramBounds[["alphaLB"]],
##
           myshape = paramBounds[["myshapeLB"]]), upper = c(L1 = paramBounds[["L1UB"]],
##
##
           alpha = paramBounds[["alphaUB"]], myshape = paramBounds[["myshapeUB"]]),
##
       control = config$control)
##
   Coefficients:
##
            T.1
                     alpha
                               myshape
##
   9.02274359 -0.08485842
                            3.88199642
##
## Log-likelihood: -394.37
## Best method: bobyqa
##
## $H2$groupB
##
## Call:
   bbmle::mle2(minuslog1 = response ~ dweibull(shape = myshape,
##
       scale = MeanLoad(L1, L1, alpha, HI)/gamma(1 + (1/myshape))),
##
       start = start, method = config$method, optimizer = config$optimizer,
##
       data = data, lower = c(L1 = paramBounds[["L1LB"]], alpha = paramBounds[["alphaLB"]],
##
##
           myshape = paramBounds[["myshapeLB"]]), upper = c(L1 = paramBounds[["L1UB"]],
           alpha = paramBounds[["alphaUB"]], myshape = paramBounds[["myshapeUB"]]),
##
       control = config$control)
##
##
  Coefficients:
##
##
           L1
                   alpha
                            myshape
##
   8.9559551 -0.2250568 4.4073411
##
## Log-likelihood: -385.11
## Best method: L-BFGS-B
##
##
## $H3
## $H3$groupA
##
## Call:
   bbmle::mle2(minuslog1 = response ~ dweibull(shape = myshape,
##
       scale = MeanLoad(L1, L2, alpha, HI)/gamma(1 + (1/myshape))),
##
       start = start, method = config$method, optimizer = config$optimizer,
##
       data = data, lower = c(L1 = paramBounds[["L1LB"]], L2 = paramBounds[["L2LB"]],
           alpha = paramBounds[["alphaLB"]], myshape = paramBounds[["myshapeLB"]]),
##
##
       upper = c(L1 = paramBounds[["L1UB"]], L2 = paramBounds[["L2UB"]],
           alpha = paramBounds[["alphaUB"]], myshape = paramBounds[["myshapeUB"]]),
##
```

```
##
       control = config$control)
##
  Coefficients:
##
##
                      L2
           T.1
                              alpha
                                        myshape
##
   9.3434134 8.4473983 -0.1819505 3.9233537
##
## Log-likelihood: -392.91
## Best method: bobyqa
##
## $H3$groupB
##
## Call:
## bbmle::mle2(minuslogl = response ~ dweibull(shape = myshape,
       scale = MeanLoad(L1, L2, alpha, HI)/gamma(1 + (1/myshape))),
##
##
       start = start, method = config$method, optimizer = config$optimizer,
       data = data, lower = c(L1 = paramBounds[["L1LB"]], L2 = paramBounds[["L2LB"]],
##
##
           alpha = paramBounds[["alphaLB"]], myshape = paramBounds[["myshapeLB"]]),
       upper = c(L1 = paramBounds[["L1UB"]], L2 = paramBounds[["L2UB"]],
##
##
           alpha = paramBounds[["alphaUB"]], myshape = paramBounds[["myshapeUB"]]),
       control = config$control)
##
##
## Coefficients:
##
           T.1
                      L2
                              alpha
                                        myshape
   8.5687035 9.1829740 -0.2393737 4.4256067
##
##
## Log-likelihood: -384.13
## Best method: bobyqa
plot_WL_Sex<- bananaPlot(mod = fitWL_Sex$H3,</pre>
             data = Field,
             response = "WL",
             group = "Sex") +
    scale_fill_manual(values = c("blue", "red")) +
  scale_color_manual(values = c("blue", "red")) +
  theme_bw()
## Scale for 'fill' is already present. Adding another scale for 'fill', which
## will replace the existing scale.
## Scale for 'colour' is already present. Adding another scale for 'colour',
## which will replace the existing scale.
plot_WL_Sex
```



"Testing H1 no alpha vs alpha" dLL dDF pvalue 1 1.23 1 0.116121 [1] "Testing H2 groupA no alpha vs alpha" dLL dDF pvalue 1 0.2 1 0.5246522 [1] "Testing H2 groupB no alpha vs alpha" dLL dDF pvalue 1 1.34 1 0.1011852 [1] "Testing H3 groupA no alpha vs alpha" dLL dDF pvalue 1 0.75 1 0.2210683 [1] "Testing H3 groupB no alpha vs alpha" dLL dDF pvalue 1 1.56 1 0.07695409 [1] "Testing H1 vs H0" dLL dDF pvalue 1 0.1 1 0.6624668 [1] "Testing H2 vs H0" dLL dDF pvalue 1 1.38 3 0.4306275 [1] "Testing H3 vs H1" dLL dDF pvalue 1 0.18242 1 0.1140242 [1] "Testing H3 vs H2" dLL dDF pvalue 1 0.18242 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.18242 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.18242 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.18242 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.182422 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.182422 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.182422 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.182422 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.182422 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.182422 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.182422 1 0.1140242 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.182422 1 0.1140242 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.182422 1 0.1140242 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.182422 1 0.1140242 1 0.1140242 1 "Testing H3 vs H2" dLL dDF pvalue 1 0.182422 1 0.114024

H0: the expected load for the subspecies and between 2 groups is the same

H1: the mean load across 2 groups is the same, but can differ across subspecies

H2: the mean load across subspecies is the same, but can differ between the 2 groups

H3: the mean load can differ both across subspecies and between 2 groups

П3:	tne mea	an 10aq	can dii	ier boti	n across	s subspe	ecies an	a betw	een z g	roups			
Fie	ld\$Sex	<- as.	factor	(Field	\$Sex)								
d	eld1 <- krop_na( eld\$MC.E	MC.Eim	eria)										
##	[1]	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
##	[13]	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
##	[25]	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
##	[37]	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
##	[49]	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
##	[61]	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
##	[73]	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
##	[85]	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
##	Г971	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA

```
## [109]
           NA
                 NA
                       NA
                             NA
                                   NA
                                         NA
                                               NA
                                                     NA
                                                           NA
                                                                 NA
                                                                       NA
                                                                             NA
## [121]
                             NΑ
                                   NΑ
                                         NΑ
                                               NΑ
                                                     NΑ
                                                           NΑ
                                                                 NΑ
                                                                       NΑ
                                                                             NΑ
           NA
                 NA
                       NA
                                                         TRUE FALSE FALSE FALSE
## [133]
           NA
                 NA
                       NA
                             NA
                                   NA
                                         NA
                                               NA
                                                   TRUE
## [145] FALSE TRUE
                    TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## [157]
         TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
              TRUE
                    TRUE
                          TRUE FALSE TRUE
                                            TRUE
                                                  TRUE
## [169]
         TRUE
                                                        TRUE FALSE
                    TRUE
                           TRUE
                                            TRUE TRUE FALSE FALSE FALSE FALSE
## [181]
         TRUE
              TRUE
                                TRUE
                                       TRUE
## [193] TRUE TRUE TRUE
                           TRUE
                                                                     TRUE TRUE
                                 TRUE FALSE FALSE FALSE TRUE TRUE
## [205] FALSE FALSE FALSE
                           TRUE TRUE FALSE FALSE FALSE FALSE
                                                                       NA FALSE
## [217] FALSE FALSE FALSE
                             NA FALSE FALSE FALSE
                                                  TRUE FALSE
                                                                 NA
                                                                     TRUE FALSE
## [229] TRUE
              TRUE FALSE
                             NA FALSE FALSE FALSE
                                                   TRUE FALSE FALSE
                                                                     TRUE TRUE
## [241] FALSE
              TRUE TRUE
                           TRUE TRUE TRUE FALSE TRUE
                                                         TRUE
                                                              TRUE
                                                                     TRUE TRUE
## [253]
           NA
               TRUE FALSE
                             NA FALSE FALSE TRUE FALSE
                                                         TRUE
                                                              TRUE
                                                                     TRUE FALSE
## [265] FALSE TRUE
                       NA FALSE FALSE FALSE FALSE
                                                         TRUE FALSE FALSE FALSE
## [277] FALSE FALSE FALSE
                           TRUE
                                   NA FALSE
                                             TRUE FALSE FALSE FALSE FALSE
## [289] FALSE FALSE TRUE
                           TRUE FALSE
                                       TRUE
                                             TRUE TRUE
                                                           NA
                                                               TRUE
                                                                     TRUE FALSE
## [301] FALSE TRUE TRUE FALSE
                                 TRUE
                                             TRUE FALSE FALSE
                                                               TRUE
                                                                    TRUE TRUE
                                         NA
## [313] TRUE FALSE FALSE
                           TRUE
                                 TRUE
                                       TRUE
                                             TRUE TRUE
                                                        TRUE TRUE FALSE
## [325] TRUE TRUE TRUE
                             NA TRUE TRUE
                                            TRUE FALSE TRUE FALSE TRUE
parasiteLoad::getParamBounds("weibull", data = Field, response = "WL")
       L1start
                       L1LB
                                    L1UB
                                              L2start
                                                              L2LB
                                                                           L2UB
##
   9.306222052 0.000000001 15.938381859 9.306222052 0.000000001 15.938381859
     alphaStart
                    alphaLB
                                 alphaUB myshapeStart
                                                         myshapeLB
                                                                      myshapeUB
   0.000000000 -5.000000000 \ \ 5.0000000000 \ \ 1.000000000 \ \ 0.000000001 \ \ 5.000000000
speparam <- c(L1start = 10,</pre>
                    L1LB = 1e-9,
                    L1UB = 20,
                    L2start = 10,
                    L2LB = 1e-9,
                    L2UB = 20,
                     alphaStart = 0, alphaLB = -5, alphaUB = 5,
                    myshapeStart = 1, myshapeLB = 1e-9, myshapeUB = 5)
##A11
fitWL_Sex <- parasiteLoad::analyse(data = Field,</pre>
                       response = "WL",
                       model = "weibull",
                       group = "Sex")
## [1] "Analysing data for response: WL"
## [1] "Fit for the response: WL"
## [1] "Fitting for all"
## [1] "Fitting model basic without alpha"
## [1] "Did converge"
## [1] "Fitting model basic with alpha"
## [1] "Did converge"
## [1] "Fitting model advanced without alpha"
## [1] "Did converge"
## [1] "Fitting model advanced with alpha"
## [1] "Did converge"
## [1] "Fitting for groupA : F"
## [1] "Fitting model basic without alpha"
```

```
## [1] "Did converge"
## [1] "Fitting model basic with alpha"
## [1] "Did converge"
## [1] "Fitting model advanced without alpha"
## [1] "Did converge"
## [1] "Fitting model advanced with alpha"
## [1] "Did converge"
## [1] "Fitting for groupB : M"
## [1] "Fitting model basic without alpha"
## [1] "Did converge"
## [1] "Fitting model basic with alpha"
## [1] "Did converge"
## [1] "Fitting model advanced without alpha"
## [1] "Did converge"
## [1] "Fitting model advanced with alpha"
## [1] "Did converge"
## [1] "Testing HO no alpha vs alpha"
##
      dLL dDF
                 pvalue
            1 0.1283225
## 1 1.16
## [1] "Testing H1 no alpha vs alpha"
##
      dLL dDF
                pvalue
## 1 1.23
            1 0.116121
## [1] "Testing H2 groupA no alpha vs alpha"
   dLL dDF
                pvalue
## 1 0.2
           1 0.5246522
## [1] "Testing H2 groupB no alpha vs alpha"
     dLL dDF
                 pvalue
## 1 1.34
           1 0.1011852
## [1] "Testing H3 groupA no alpha vs alpha"
##
      dLL dDF
                 pvalue
## 1 0.75
            1 0.2210683
## [1] "Testing H3 groupB no alpha vs alpha"
      dLL dDF
                  pvalue
            1 0.07695409
## 1 1.56
## [1] "Testing H1 vs H0"
   dLL dDF
                pvalue
## 1 0.1
           1 0.6624668
## [1] "Testing H2 vs H0"
##
      dLL dDF
                 pvalue
## 1 1.38
            3 0.4306275
## [1] "Testing H3 vs H1"
     dLL dDF
                 pvalue
## 1 3.72
            4 0.1140242
## [1] "Testing H3 vs H2"
     dLL dDF
                  pvalue
## 1 2.44
            2 0.08709369
fitWL_Sex
## $HO
##
## bbmle::mle2(minuslog1 = response ~ dweibull(shape = myshape,
       scale = MeanLoad(L1, L1, alpha, HI)/gamma(1 + (1/myshape))),
##
       start = start, method = config$method, optimizer = config$optimizer,
```

```
data = data, lower = c(L1 = paramBounds[["L1LB"]], alpha = paramBounds[["alphaLB"]],
##
##
           myshape = paramBounds[["myshapeLB"]]), upper = c(L1 = paramBounds[["L1UB"]],
           alpha = paramBounds[["alphaUB"]], myshape = paramBounds[["myshapeUB"]]),
##
       control = config$control)
##
##
  Coefficients:
##
                   alpha
                            myshape
   9.0079495 -0.1450148 4.1118351
##
##
## Log-likelihood: -780.86
## Best method: L-BFGS-B
##
## $H1
##
## Call:
   bbmle::mle2(minuslog1 = response ~ dweibull(shape = myshape,
##
       scale = MeanLoad(L1, L2, alpha, HI)/gamma(1 + (1/myshape))),
       start = start, method = config$method, optimizer = config$optimizer,
##
       data = data, lower = c(L1 = paramBounds[["L1LB"]], L2 = paramBounds[["L2LB"]],
##
           alpha = paramBounds[["alphaLB"]], myshape = paramBounds[["myshapeLB"]]),
##
##
       upper = c(L1 = paramBounds[["L1UB"]], L2 = paramBounds[["L2UB"]],
##
           alpha = paramBounds[["alphaUB"]], myshape = paramBounds[["myshapeUB"]]),
       control = config$control)
##
##
  Coefficients:
                      L2
                              alpha
                                        myshape
##
   9.0783810 8.9308463 -0.1526132 4.1153069
##
## Log-likelihood: -780.77
## Best method: bobyqa
##
## $H2
## $H2$groupA
##
## Call:
## bbmle::mle2(minuslog1 = response ~ dweibull(shape = myshape,
##
       scale = MeanLoad(L1, L1, alpha, HI)/gamma(1 + (1/myshape))),
##
       start = start, method = config$method, optimizer = config$optimizer,
       data = data, lower = c(L1 = paramBounds[["L1LB"]], alpha = paramBounds[["alphaLB"]],
##
           myshape = paramBounds[["myshapeLB"]]), upper = c(L1 = paramBounds[["L1UB"]],
##
           alpha = paramBounds[["alphaUB"]], myshape = paramBounds[["myshapeUB"]]),
##
##
       control = config$control)
##
   Coefficients:
##
##
                     alpha
                               myshape
   9.02274359 -0.08485842 3.88199642
##
##
## Log-likelihood: -394.37
## Best method: bobyqa
## $H2$groupB
##
## Call:
## bbmle::mle2(minuslog1 = response ~ dweibul1(shape = myshape,
```

```
scale = MeanLoad(L1, L1, alpha, HI)/gamma(1 + (1/myshape))),
##
##
       start = start, method = config$method, optimizer = config$optimizer,
       data = data, lower = c(L1 = paramBounds[["L1LB"]], alpha = paramBounds[["alphaLB"]],
##
           myshape = paramBounds[["myshapeLB"]]), upper = c(L1 = paramBounds[["L1UB"]],
##
           alpha = paramBounds[["alphaUB"]], myshape = paramBounds[["myshapeUB"]]),
##
       control = config$control)
##
##
##
  Coefficients:
##
           T.1
                   alpha
                            myshape
   8.9559551 -0.2250568
                         4.4073411
##
## Log-likelihood: -385.11
## Best method: L-BFGS-B
##
##
## $H3
## $H3$groupA
##
## Call:
## bbmle::mle2(minuslog1 = response ~ dweibul1(shape = myshape,
##
       scale = MeanLoad(L1, L2, alpha, HI)/gamma(1 + (1/myshape))),
##
       start = start, method = config$method, optimizer = config$optimizer,
       data = data, lower = c(L1 = paramBounds[["L1LB"]], L2 = paramBounds[["L2LB"]],
##
           alpha = paramBounds[["alphaLB"]], myshape = paramBounds[["myshapeLB"]]),
##
       upper = c(L1 = paramBounds[["L1UB"]], L2 = paramBounds[["L2UB"]],
##
##
           alpha = paramBounds[["alphaUB"]], myshape = paramBounds[["myshapeUB"]]),
##
       control = config$control)
##
  Coefficients:
##
##
           L1
                      L2
                              alpha
                                        myshape
   9.3434134 8.4473983 -0.1819505 3.9233537
##
##
## Log-likelihood: -392.91
## Best method: bobyqa
## $H3$groupB
##
## Call:
## bbmle::mle2(minuslog1 = response ~ dweibull(shape = myshape,
       scale = MeanLoad(L1, L2, alpha, HI)/gamma(1 + (1/myshape))),
##
##
       start = start, method = config$method, optimizer = config$optimizer,
       data = data, lower = c(L1 = paramBounds[["L1LB"]], L2 = paramBounds[["L2LB"]],
##
           alpha = paramBounds[["alphaLB"]], myshape = paramBounds[["myshapeLB"]]),
##
##
       upper = c(L1 = paramBounds[["L1UB"]], L2 = paramBounds[["L2UB"]],
           alpha = paramBounds[["alphaUB"]], myshape = paramBounds[["myshapeUB"]]),
##
       control = config$control)
##
##
  Coefficients:
##
##
                      L2
                              alpha
                                        myshape
   8.5687035 9.1829740 -0.2393737
##
                                     4.4256067
##
## Log-likelihood: -384.13
## Best method: bobyga
```

## Scale for 'fill' is already present. Adding another scale for 'fill', which
## will replace the existing scale.

## Scale for 'colour' is already present. Adding another scale for 'colour',
## which will replace the existing scale.

plot\_WL\_Sex

