

# HR\_Va\_h2\_G1\_2023

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```
#load pHRkages
```

```
library(lme4)
```

```
## Loading required package: Matrix
```

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
```

```
## v dplyr      1.1.4      v readr      2.1.5
```

```
## v forcats   1.0.0      v stringr   1.5.1
```

```
## v ggplot2    3.4.4      v tibble    3.2.1
```

```
## v lubridate  1.9.3      v tidyr     1.3.0
```

```
## v purrr      1.0.2
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x tidyr::expand() masks Matrix::expand()
```

```
## x dplyr::filter() masks stats::filter()
```

```
## x dplyr::lag()    masks stats::lag()
```

```
## x tidyr::pack()   masks Matrix::pack()
```

```
## x tidyr::unpack() masks Matrix::unpack()
```

```
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(lmerTest)
```

```
##
```

```
## Attaching package: 'lmerTest'
```

```
##
```

```
## The following object is masked from 'package:lme4':
```

```
##
```

```
##      lmer
```

```
##
```

```
## The following object is masked from 'package:stats':
```

```
##
```

```
##      step
```

```
library(car)
```

```
## Loading required package: carData
```

```
##
```

```
## Attaching package: 'car'
```

```
##
```

```
## The following object is masked from 'package:dplyr':
```

```
##
```

```
##      recode
```

```
##
```

```

## The following object is masked from 'package:purrr':
##
##      some

#####2023#####

Read in the data:
HR_23 <- read_csv(here::here("data_sheets", "compiled_sheets", "HR_mastersheet_Fitness-mains_2023.csv"))

## Rows: 781 Columns: 51
## -- Column specification -----
## Delimiter: ","
## chr   (6): Gen, Donor, Recipient, sample_ID_SEG, sample_ID, SegPos
## dbl  (35): Year, Transect, Sequence, Plant_ID, days_sow2flower, days_plant2f...
## lgl   (1): F_plant
## date  (9): Sow_Date, Plant_Date, FFD, LFD, F_Num_01, F_Num_02, F_Num_03, pho...
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
HR_23_fit <- HR_23 %>%
  filter(Gen == "G1")

HR_22_23_full <- read_csv(here::here("data_sheets", "compiled_sheets", "HR_22_23_full.csv"))

## Warning: One or more parsing issues, call `problems()` on your data frame for details,
## e.g.:
##   dat <- vroom(...)
##   problems(dat)

## Rows: 7684 Columns: 67
## -- Column specification -----
## Delimiter: ","
## chr   (9): Gen, Donor, Recipient, sample_ID_SEG, sample_ID, SegPos, Block, R...
## dbl  (45): Year, Transect, Sequence, Plant_ID, days_sow2flower, days_plant2f...
## lgl   (4): F_plant, F_Num_03, Rep_FitP, any_FitP
## date  (9): Sow_Date, Plant_Date, FFD, LFD, F_Num_01, F_Num_02, photo_date, p...
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
# Calculate the mean surv_to_flower for each group in AC_22_23_full and add it as a new column
HR_22_23_full <- HR_22_23_full %>%
  group_by(Year, Gen, Transect, Sequence, Donor, Recipient, SegPos) %>%
  mutate(prop_surv_to_flower = mean(surv_to_flower, na.rm = TRUE)) %>%
  ungroup() # Remove grouping

HR_prop_sample <- HR_22_23_full %>%
  select(c(Year, Gen, Transect, Sequence, Donor, Recipient, SegPos, prop_surv_to_flower)) %>%
  distinct()

HR_23_fit <- HR_23_fit %>%
  left_join(HR_prop_sample %>% select(Year, Gen, Transect, Sequence, Donor, Recipient, SegPos, prop_surv_to_flower),
    by = c("Year", "Gen", "Transect", "Sequence", "Donor", "Recipient"))

## Warning in left_join(., HR_prop_sample %>% select(Year, Gen, Transect, Sequence, : Detected an unexp
## i Row 102 of `x` matches multiple rows in `y`.

```

```

## i Row 195 of `y` matches multiple rows in `x`.
## i If a many-to-many relationship is expected, set `relationship =
## "many-to-many"` to silence this warning.

# Create the new column
HR_23_fit <- HR_23_fit %>%
  mutate(est_fitness = prop_surv_to_flower * est_fecundity)

#log transform traits that need it
HR_23_fit$skel_dryweight_mg_SEG <- log(HR_23_fit$skel_dryweight_mg_SEG)
HR_23_fit$msm_all <- sqrt(HR_23_fit$msm_all + 1)
HR_23_fit$SLA_SEG <- log(HR_23_fit$SLA_SEG)
HR_23_fit$est_fitness <- sqrt(HR_23_fit$est_fitness)

#mean center the traits of interest

traits <- c("corolla_diam_mm_SEG", "skel_dryweight_mg_SEG", "fl_duration", "est_fecundity", "msm_all",

# Mean center eHRh trait
for (trait in traits) {
  trait_mean <- mean(HR_23_fit[[trait]], na.rm = TRUE)
  HR_23_fit[[paste0(trait, "_centered")]] <- HR_23_fit[[trait]] - trait_mean
}

# Create the mixed model for corolla area

#corolla_model <- lmer(corolla_diam_mm_SEG_centered ~ (1 | Recipient) + (1 | Donor), data = HR_23_fit)

# Create the mixed model for skeleton weight, with skeleton weight log transformed
corolla_model <- lmer((corolla_diam_mm_SEG_centered) ~ (1 | Transect) + (1 | Donor), data = HR_23_fit)

rand(corolla_model)

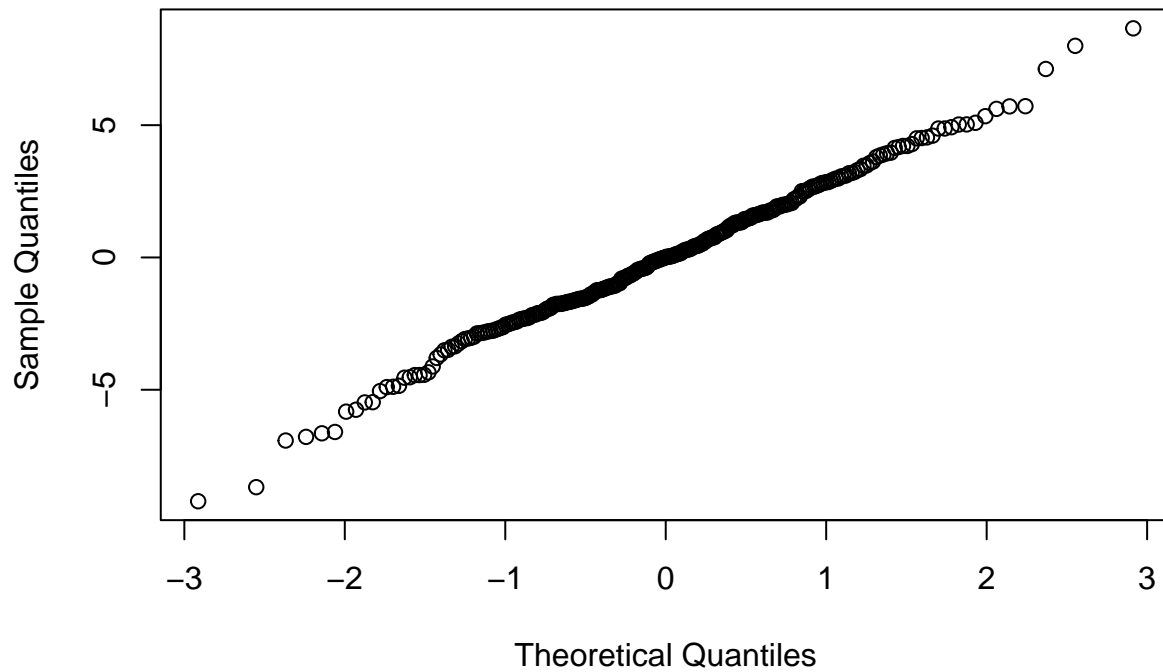
## ANOVA-like table for random-effects: Single term deletions
##
## Model:
## (corolla_diam_mm_SEG_centered) ~ (1 | Transect) + (1 | Donor)
##               npar logLik   AIC    LRT Df Pr(>Chisq)
## <none>          4 -698.11 1404.2
## (1 | Transect)   3 -702.02 1410.0  7.8165  1  0.005177 **
## (1 | Donor)      3 -698.12 1402.2  0.0205  1  0.886042
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# ExtrHRT residuals from the model
residuals <- resid(corolla_model)

# Q-Q plot for normality
qqnorm(residuals) #looks good

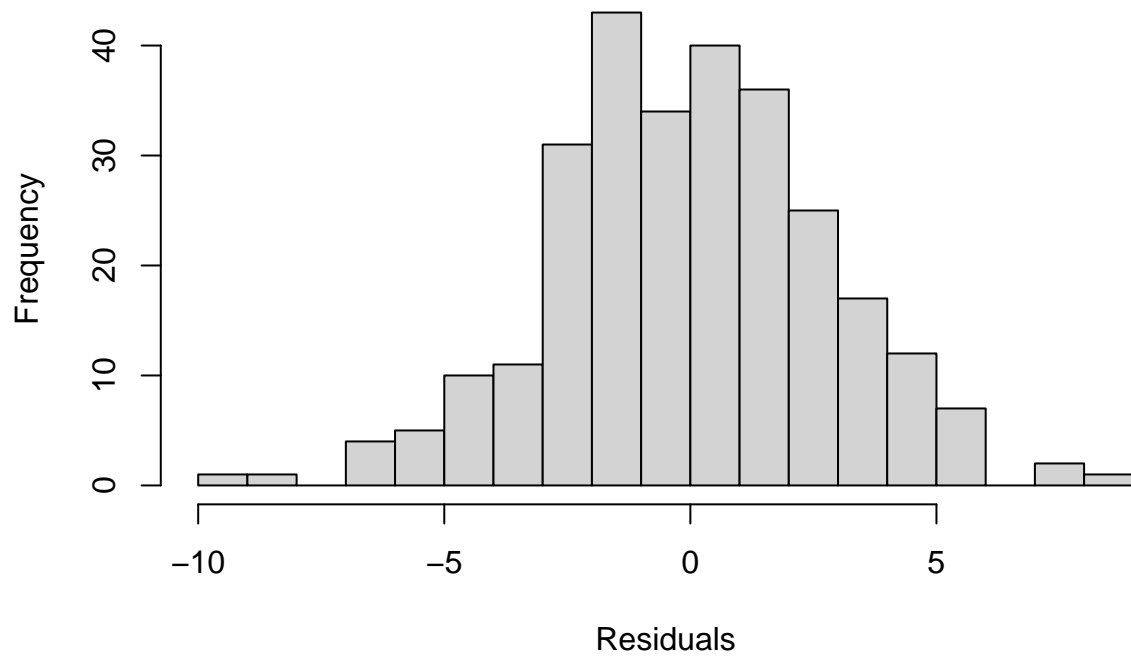
```

## Normal Q-Q Plot



```
# Histogram for normality
hist(residuals, breaks = 20, main = "Histogram of Residuals", xlab = "Residuals") #normal
```

## Histogram of Residuals



```
# Create the mixed model for skeleton weight, with skeleton weight log transformed
#skel_model <- lmer((skel_dryweight_mg_SEG_centered) ~ (1 | Recipient) + (1 | Donor), data = HR_23_fit)
```

```
# Create the mixed model for skeleton weight, with skeleton weight log transformed
skel_model <- lmer((skel_dryweight_mg_SEG_centered) ~ (1 | Donor), data = HR_23_fit)
```

```
# Test the significance of the random effects
rand(skel_model)
```

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

```
##
```

```
## Model:
```

```
## (skel_dryweight_mg_SEG_centered) ~ (1 | Donor)
```

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

```
## <none>         3 -486.51 979.03
```

```
## (1 | Donor)    2 -486.54 977.07 0.043864 1      0.8341
```

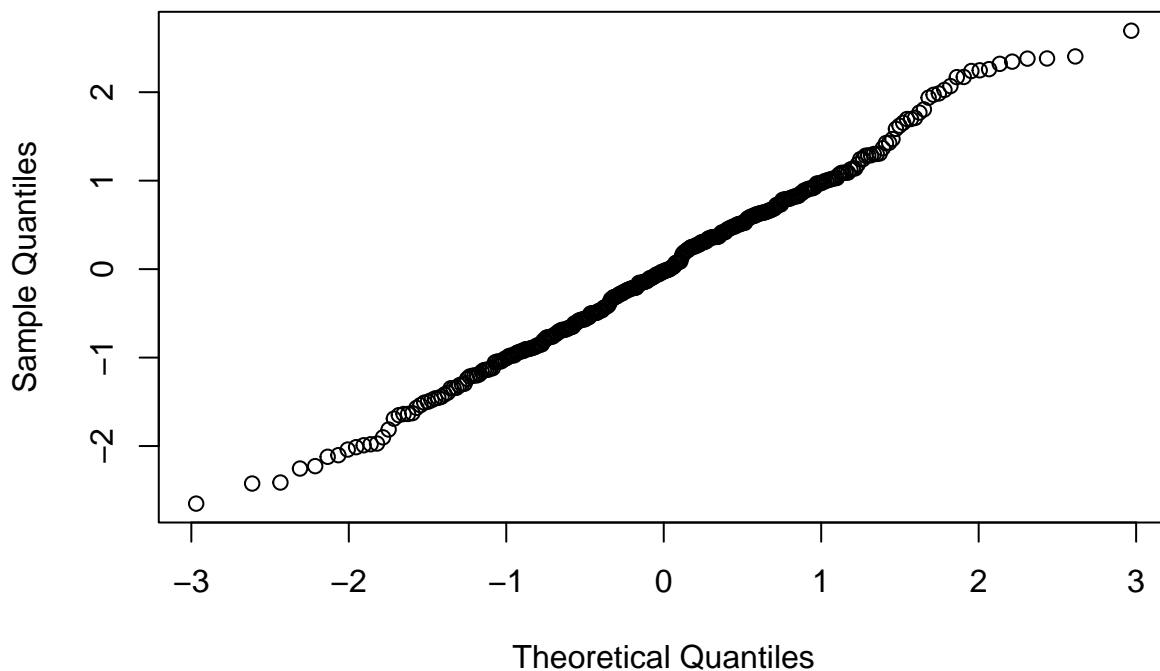
```
# ExtrHRT residuals from the model
```

```
residuals <- resid(skel_model)
```

```
# Q-Q plot for normality
```

```
qqnorm(residuals) #looks good
```

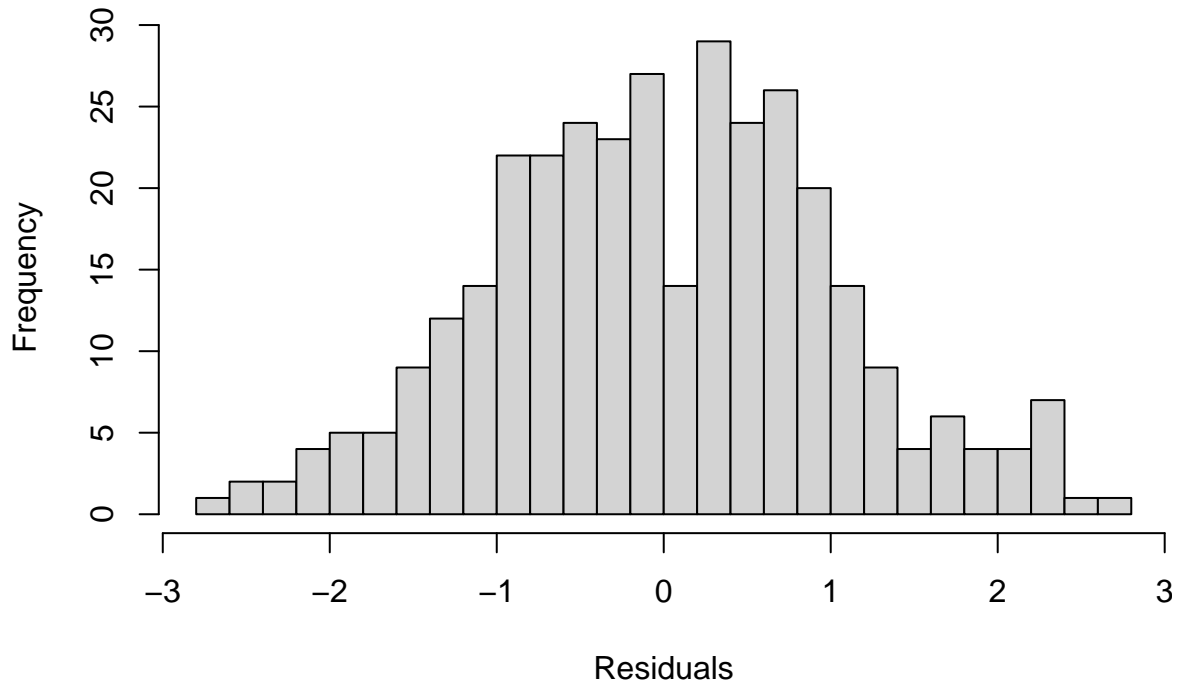
**Normal Q-Q Plot**



```
# Histogram for normality
```

```
hist(residuals, breaks = 20, main = "Histogram of Residuals", xlab = "Residuals") #normal
```

## Histogram of Residuals



```
# Create the mixed model for flowering duration
#fl_duration_model <- lmer(fl_duration_centered ~ (1 | Recipient) + (1 | Donor), data = HR_23_fit) #Singular

#fl_duration_model <- lmer(fl_duration_centered ~ (1 | Donor), data = HR_23_fit) #Singular

# Create the mixed model for estimated fecundity, sqrt transforming estimated fecundity
#est_fecundity_model <- lmer((est_fecundity_centered) ~ (1 | Recipient) + (1 | Donor), data = HR_23_fit)

# Create the mixed model for estimated fecundity, sqrt transforming estimated fecundity
#est_fecundity_model_2 <- lmer((est_fecundity_centered) ~ (1 | Donor), data = HR_23_fit) #still singular

# Create the mixed model for mean seed mass, log transformed mean seed mass
msm_model <- lmer((msm_all_centered) ~ (1 | Transect) + (1 | Recipient) + (1 | Donor), data = HR_23_fit)

## Warning in checkConv(attr("derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.00319197 (tol = 0.002, component 1)

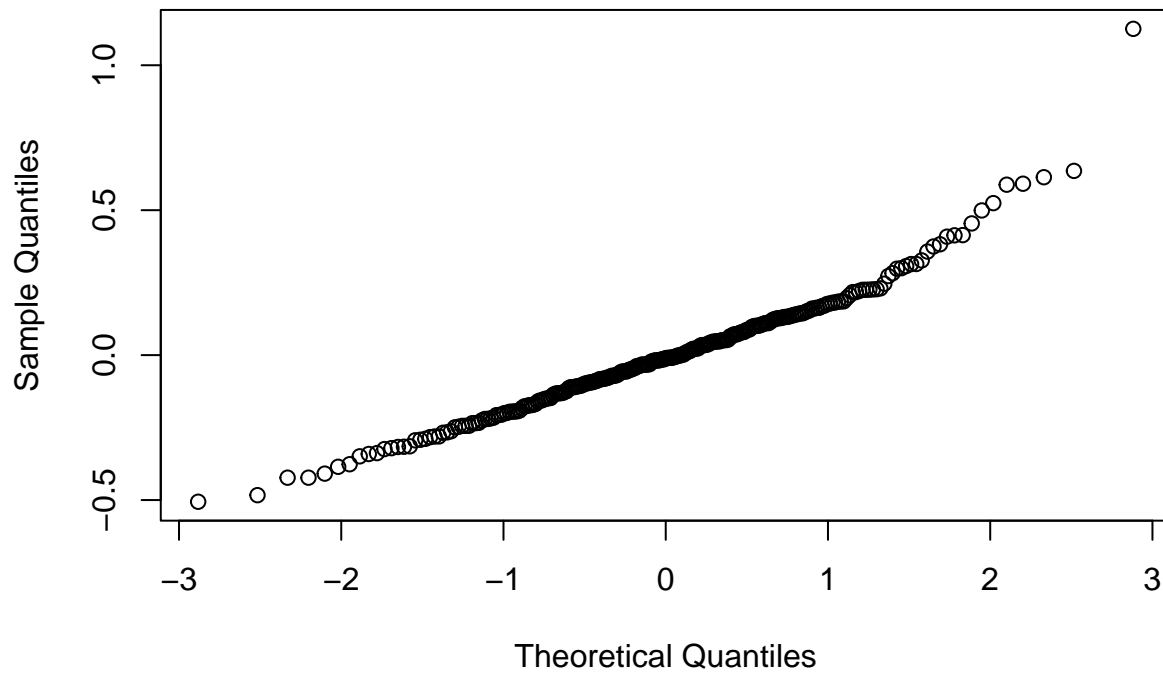
rand(msm_model)

## ANOVA-like table for random-effects: Single term deletions
##
## Model:
## (msm_all_centered) ~ (1 | Transect) + (1 | Recipient) + (1 | Donor)
##      npar  logLik   AIC    LRT Df Pr(>Chisq)
## <none>      5 -34.920 79.841
## (1 | Transect)    4 -36.815 81.630 3.7888  1  0.051596 .
## (1 | Recipient)   4 -39.030 86.060 8.2191  1  0.004145 **
## (1 | Donor)      4 -34.924 77.848 0.0074  1  0.931291
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# ExtrHRT residuals from the model
residuals <- resid(msm_model)

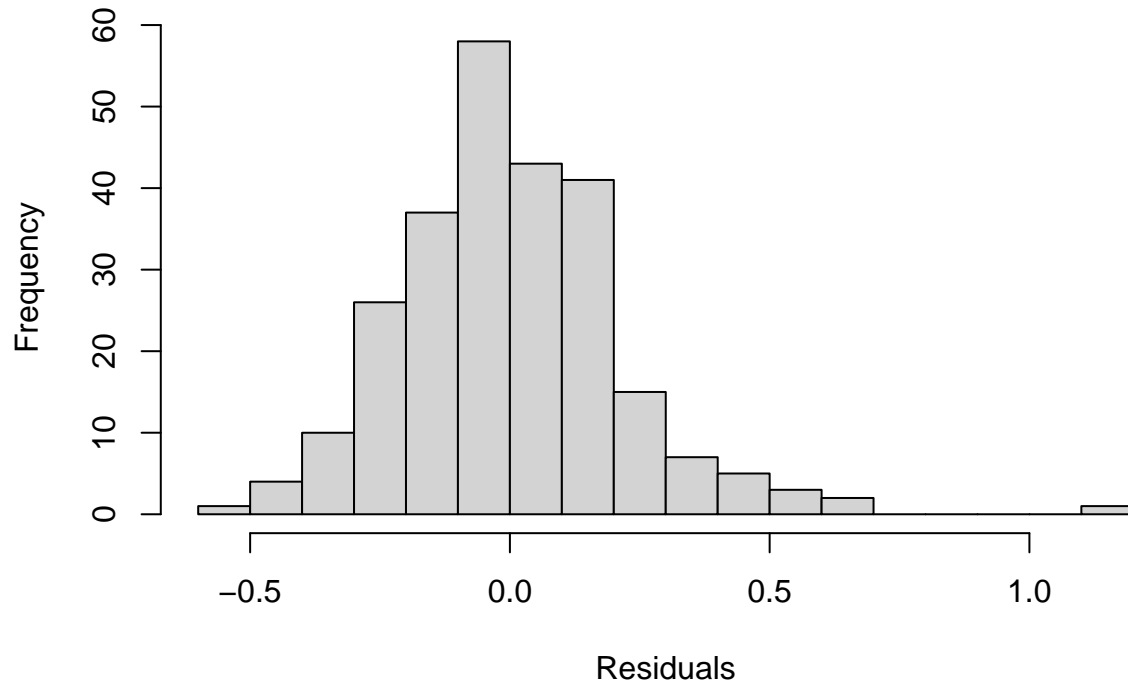
# Q-Q plot for normality
qqnorm(residuals) #good enough
```

Normal Q-Q Plot



```
# Histogram for normality
hist(residuals, breaks = 20, main = "Histogram of Residuals", xlab = "Residuals") #normal-ish
```

## Histogram of Residuals



```
# Create the mixed model for SLA
#SLA_model <- lmer((SLA_SEG_centered) ~ (1 | Recipient) + (1 | Donor), data = HR_23_fit) #singular

#SLA_model <- lmer((SLA_SEG_centered) ~ (1 | Donor), data = HR_23_fit) #Singular

# Create the mixed model for mean seed mass, log transformed mean seed mass
#LMA_model <- lmer(LMA_SEG_centered ~ (1 | Donor), data = HR_23_fit) #Singular

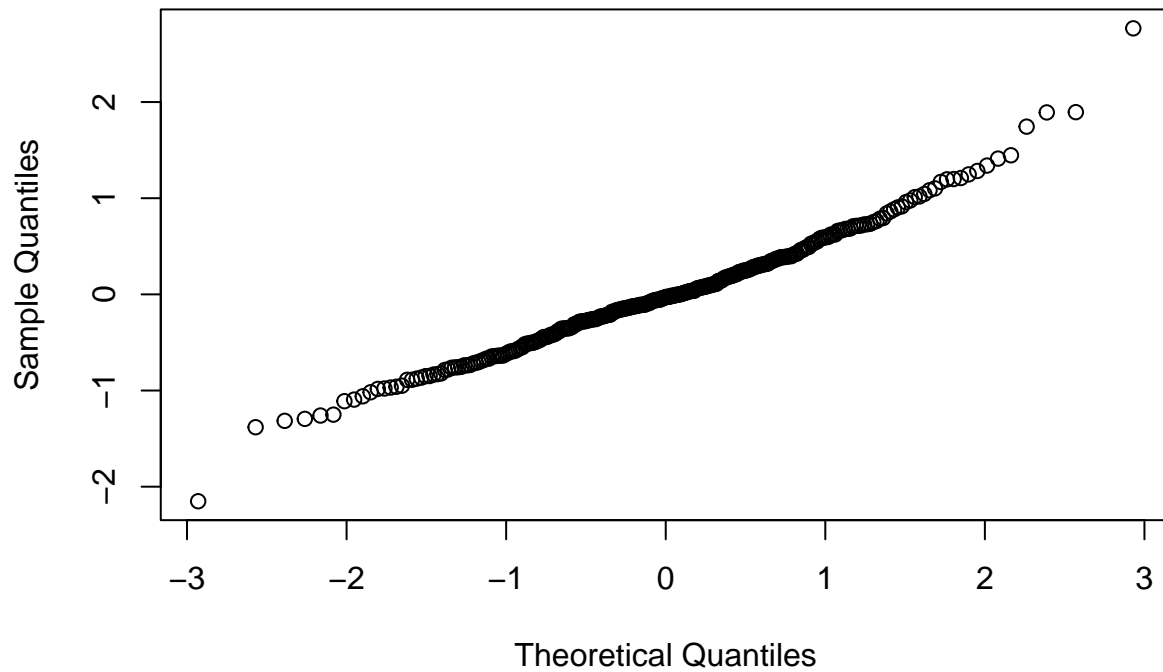
# Create the mixed model for d13C
d13C_model <- lmer(d13C_SEG_centered ~ (1 | Transect) + (1 | Recipient) + (1 | Donor), data = HR_23_fit)

# ExtrHRT residuals from the model
residuals <- resid(d13C_model)

# Q-Q plot for normality
qqnorm(residuals) #good enough
```

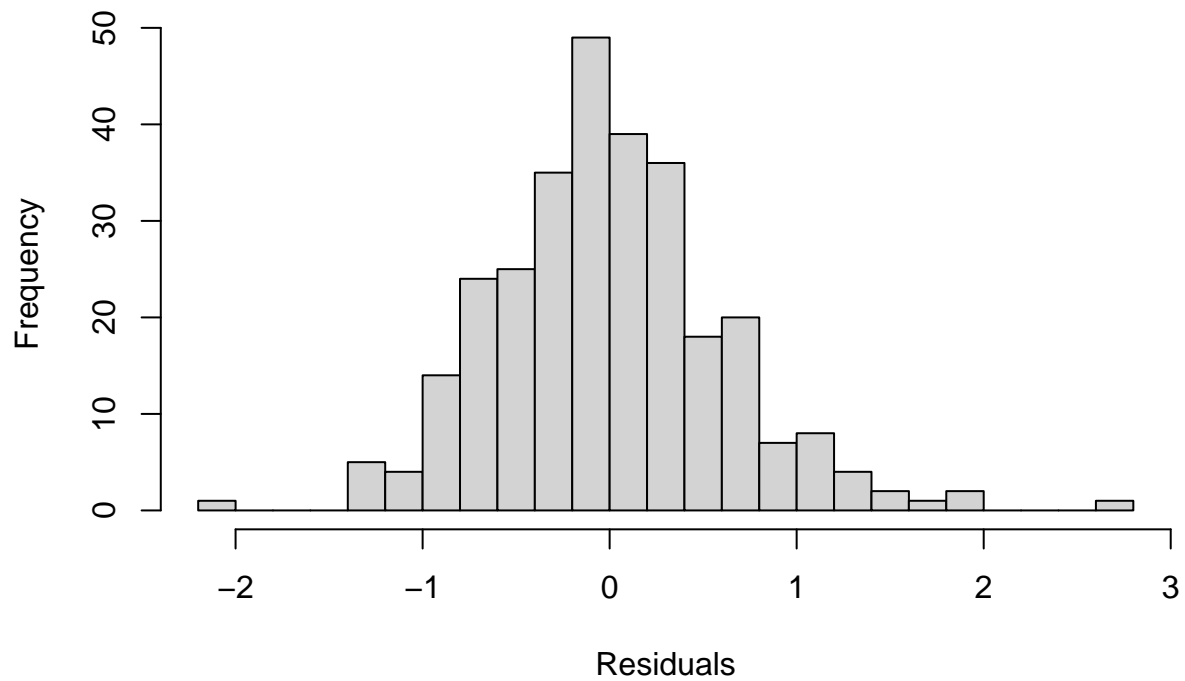


## Normal Q-Q Plot



```
# Histogram for normality  
hist(residuals, breaks = 20, main = "Histogram of Residuals", xlab = "Residuals") #normal-ish
```

## Histogram of Residuals



```
rand(d13C_model)
```

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

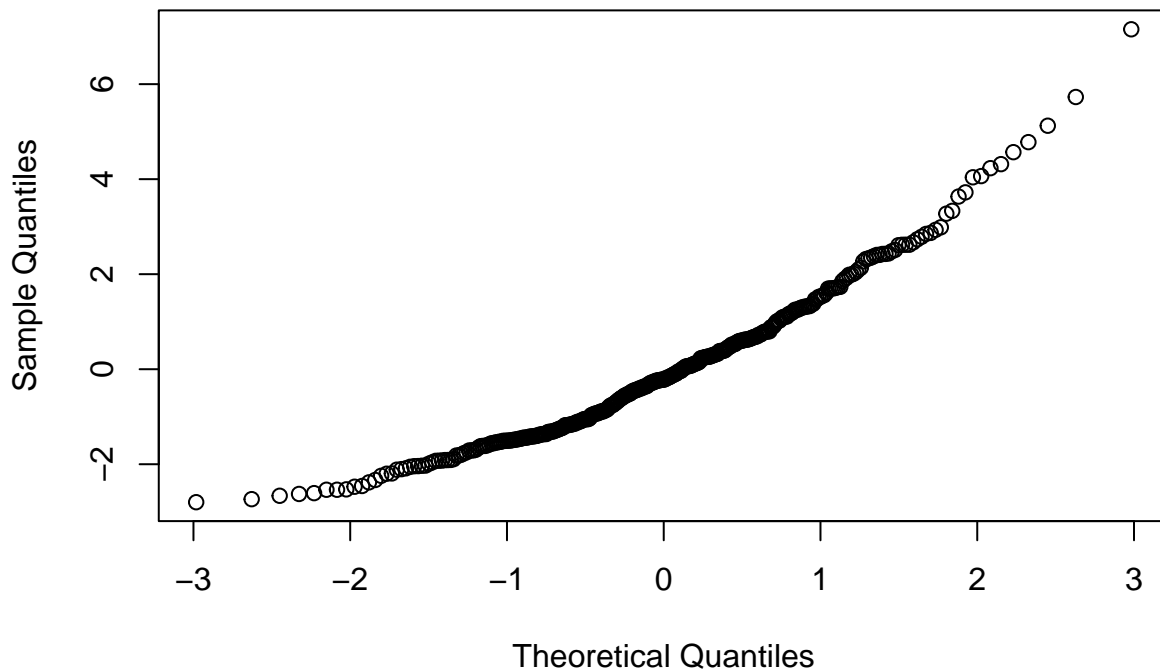
```
##
## Model:
## d13C_SEG_centered ~ (1 | Transect) + (1 | Recipient) + (1 | Donor)
##               npar  logLik   AIC    LRT Df Pr(>Chisq)
## <none>          5 -328.39 666.77
## (1 | Transect)   4 -331.66 671.33 6.5543  1    0.01046 *
## (1 | Recipient)  4 -328.87 665.73 0.9571  1    0.32793
## (1 | Donor)      4 -330.23 668.46 3.6848  1    0.05491 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# Create the mixed model for d13C, log transformed mean seed mass
est_fitness_model <- lmer(est_fitness_centered ~ (1|Transect) + (1 | Recipient) + (1 | Donor), data = HI)

# Extract residuals from the model
residuals <- resid(est_fitness_model)

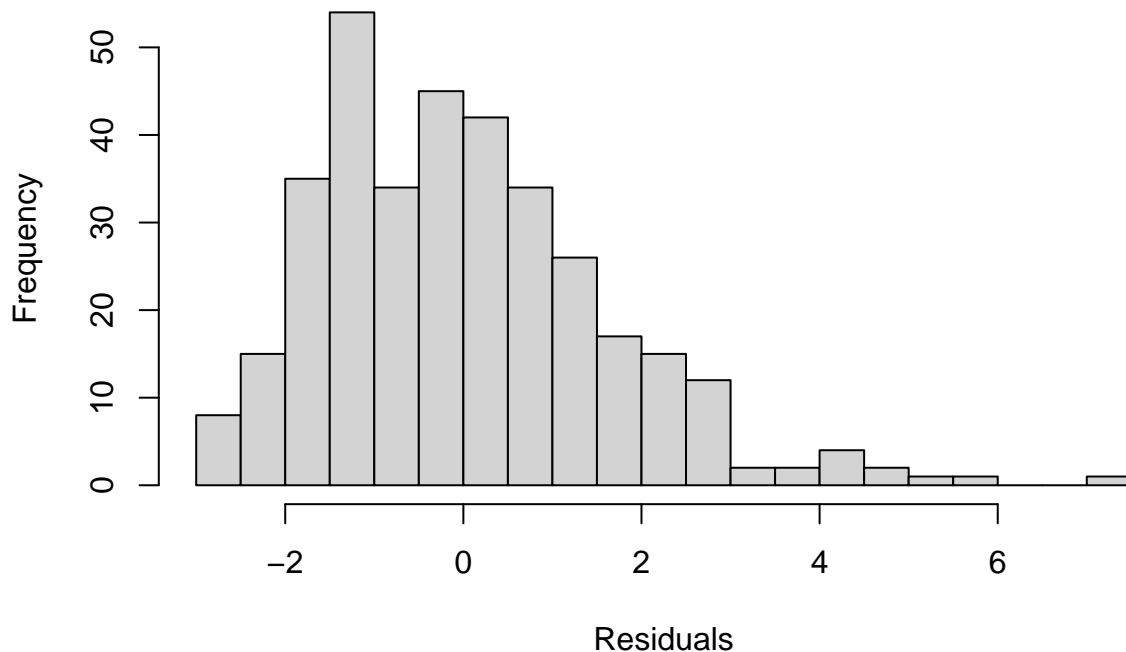
# Q-Q plot for normality
qqnorm(residuals) #good enough
```

Normal Q-Q Plot



```
# Histogram for normality
hist(residuals, breaks = 20, main = "Histogram of Residuals", xlab = "Residuals") #normal-ish
```

## Histogram of Residuals



```
# Test the significance of the random effects
rand(est_fitness_model) #all Random effects significantly improve the models fit!
```

```
## ANOVA-like table for random-effects: Single term deletions
##
## Model:
## est_fitness_centered ~ (1 | Transect) + (1 | Recipient) + (1 | Donor)
##      npar  logLik    AIC    LRT Df Pr(>Chisq)
## <none>      5 -709.19 1428.4
## (1 | Transect)    4 -719.56 1447.1 20.7566  1  5.215e-06 ***
## (1 | Recipient)    4 -710.16 1428.3  1.9461  1    0.1630
## (1 | Donor)       4 -709.31 1426.6  0.2392  1    0.6248
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# Function to round values to a specified number of significant digits
round_df <- function(df, digits) {
  df[] <- lapply(df, function(x) if(is.numeric(x)) signif(x, digits) else x)
  return(df)
}
```

```
# Function to extract variance components and calculate required values
calculate_variances <- function(model, trait_name) {
  var_components <- as.data.frame(VarCorr(model))

  # Initialize variables
  V_mat <- NA
  V_sd_mat <- NA
  Va_mat <- NA
  V_pat <- NA
```

```

V_sd_pat <- NA
Va_pat <- NA
res_var <- NA
Vp <- NA
Vp_sd <- NA
h2 <- NA

# Check if "Recipient" is included as a random effect
if ("Recipient" %in% var_components$grp) {
  V_mat <- var_components$vcov[var_components$grp == "Recipient"]
  V_sd_mat <- sqrt(V_mat)
  Va_mat <- V_mat * 4
}

# Calculate other variance components
if ("Donor" %in% var_components$grp) {
  V_pat <- var_components$vcov[var_components$grp == "Donor"]
  V_sd_pat <- sqrt(V_pat)
  Va_pat <- V_pat * 4
}

if ("Residual" %in% var_components$grp) {
  res_var <- var_components$vcov[var_components$grp == "Residual"]
}

# Calculate total phenotypic variance and narrow-sense heritability if components are available
if (!is.na(Va_mat) & !is.na(Va_pat) & !is.na(res_var)) {
  Vp <- Va_mat + Va_pat + res_var
  Vp_sd <- sqrt(Vp)
  h2 <- Va_pat / Vp # assumed calculation
}

# Extract the number of observations
n_obs <- nobs(model)

# Create the dataframe and add the traits column
df <- data.frame(traits = trait_name, V_mat, V_sd_mat, Va_mat, V_pat, V_sd_pat, Va_pat, Vp, Vp_sd, h2)

# Round the dataframe values to four significant digits
df <- round_df(df, 4)

return(df)
}

# Calculate variances for eHRh model and add trait names
corolla_variances <- calculate_variances(corolla_model, 'corolla_diameter')
skel_variances <- calculate_variances(skel_model, "skel_biomass_mg")
#est_fecundity_variances <- calculate_variances(est_fecundity_model, "estimated_fecundity")
msm_variances <- calculate_variances(msm_model, "mean_seed_mass")
#SLA_variances <- calculate_variances(SLA_model, "SLA")
#LMA_variances <- calculate_variances(LMA_model, "LMA")
d13C_variances <- calculate_variances(d13C_model, "delta_C_13")
est_fitness_variances <- calculate_variances(est_fitness_model, "est_fitness")

```

```
# Combine the results into a single dataframe
```

```
variance_HR_2023_G1 <- rbind(  
  corolla_variances,  
  skel_variances,  
  dl3C_variances,  
  msm_variances,  
  est_fitness_variances  
)
```

```
# Print the dataframe
```

```
print(variance_HR_2023_G1)
```

```
##          traits    V_mat V_sd_mat  Va_mat    V_pat V_sd_pat  Va_pat    Vp  
## 1 corolla_diameter      NA      NA      NA 0.0433300 0.20820 0.173300    NA  
## 2  skel_biomass_mg      NA      NA      NA 0.0070000 0.08366 0.028000    NA  
## 3      delta_C_13 0.03849  0.1962 0.15400 0.0550500 0.23460 0.220200 0.8200  
## 4  mean_seed_mass 0.01848  0.1359 0.07391 0.0003819 0.01954 0.001528 0.1328  
## 5      est_fitness 0.31020  0.5570 1.24100 0.0662500 0.25740 0.265000 4.3930  
##    Vp_sd      h2 n_obs  
## 1    NA      NA   280  
## 2    NA      NA   335  
## 3 0.9055 0.26860   295  
## 4 0.3645 0.01150   253  
## 5 2.0960 0.06032   350
```

```
#Save the csv file if you want
```

```
write_csv(x = variance_HR_2023_G1, here::here("data_sheets", "compiled_sheets", "HR_Va_h2_R_2023.csv"))
```