

Scripps's Murrelet Egg Size model

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#TODO - Calculate confidence intervals - Prediction

Introduction

This document details steps taken to model Scripps's Murrelet (*Synthliboramphus scrippsi*) egg size at Santa Barbara Island within Channel Islands National Park from 2009-2017.

Load and format data

```
egg <- read_csv(here("data", "SCMU_egg_data.csv"))
```

```
## Warning: Missing column names filled in: 'X1' [1]
```

```
##
## -- Column specification -----
## cols(
##   X1 = col_double(),
##   ID = col_character(),
##   Year = col_double(),
##   ObservationDate = col_date(format = ""),
##   Observer = col_character(),
##   Plot = col_character(),
##   NestNumber = col_character(),
##   NestContents = col_character(),
##   EggOrderKnown = col_character(),
##   TrueOrder = col_logical(),
##   EggOrder = col_character(),
##   EggState = col_character(),
##   Length = col_double(),
##   Width = col_double(),
##   Size = col_double(),
##   CommentList = col_character(),
##   AdditionalComments = col_character()
## )
```

```
covars <- read_csv(here("data", "covariates", "covars.csv"))
```

```
## Warning: Missing column names filled in: 'X1' [1]
```

```
##
## -- Column specification -----
```

```
## cols(
##   X1 = col_double(),
##   Year = col_double(),
##   ANCHL = col_double(),
##   BEUTI = col_double(),
##   NPGO = col_double(),
##   ONI = col_double(),
##   PDO = col_double(),
##   SST = col_double()
## )

## join covariate data with egg data by year
df <- left_join(egg, covars, by = "Year") %>%
  filter(TrueOrder == TRUE) %>% # egg order known only
  select(Year, Observer, Plot, Size, EggOrder, ANCHL, BEUTI, NPGO, ONI, PDO, SST)
```

Null Intercept-Only Model

This model includes observer and plot as random effects, but does not include any covariates.

```
## run model
nm <- lmer(Size ~ 1 + (1 | Observer) + (1 | Plot), data = df)

## look at model output and estimates
summary(nm)
```

```
## Fixed Effects:
## coef.est  coef.se
## 1884.29    12.20
##
## Random Effects:
## Groups   Name      Std.Dev.
## Observer (Intercept) 17.22
## Plot      (Intercept) 24.49
## Residual              111.96
## ---
## number of obs: 753, groups: Observer, 27; Plot, 8
## AIC = 9263.1, DIC = 9268.5
## deviance = 9261.8
```

```
coef(nm) # these are the coefficients
```

```
## $Observer
##      (Intercept)
## AAY      1899.553
## AJB      1880.127
## AJD      1895.666
## AML      1886.089
## CAC      1876.699
## CEH      1880.914
## CEK      1880.307
## CLE      1879.914
## DLW      1874.348
## EWW      1891.898
## GRK      1882.978
```

```

## JAH      1902.680
## KMR      1885.299
## KWB      1880.009
## LAH      1883.315
## MEJ      1878.435
## MGB      1886.816
## NAG      1869.396
## PTL      1876.981
## RER      1893.489
## REW      1881.330
## SAA      1887.456
## SFC      1885.064
## SJK      1893.394
## SKT      1862.709
## SLA      1884.647
## SMC      1896.380
##
## $Plot
##      (Intercept)
## APNC      1905.622
## BH        1898.588
## BT        1850.797
## CC        1903.414
## DO        1885.298
## ESC       1882.827
## LC        1873.535
## WC        1874.258
##
## attr("class")
## [1] "coef.mer"

```

```
ranef(nm) # these are the random effects
```

```

## $Observer
##      (Intercept)
## AAY  15.2610062
## AJB  -4.1654786
## AJD  11.3739567
## AML   1.7970244
## CAC  -7.5934214
## CEH  -3.3784652
## CEK  -3.9853109
## CLE  -4.3779272
## DLW  -9.9445690
## EWW   7.6052536
## GRK  -1.3144290
## JAH  18.3876524
## KMR   1.0067101
## KWB  -4.2835217
## LAH  -0.9770916
## MEJ  -5.8572070
## MGB   2.5241208
## NAG -14.8959509
## PTL  -7.3117215
## RER   9.1963408

```

```

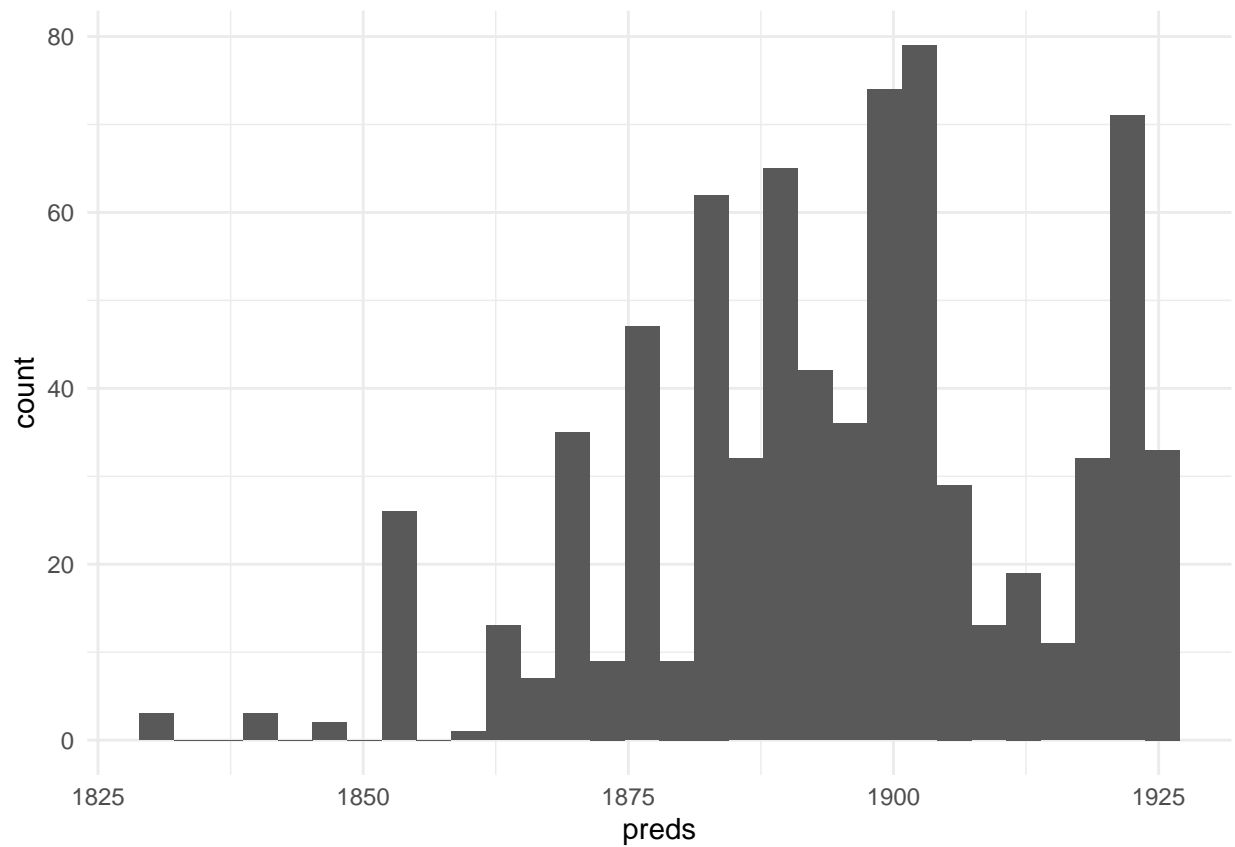
## REW -2.9625711
## SAA 3.1632622
## SFC 0.7713125
## SJK 9.1015277
## SKT -21.5831481
## SLA 0.3551011
## SMC 12.0875447
##
## $Plot
## (Intercept)
## APNC 21.329246
## BH 14.295406
## BT -33.494909
## CC 19.121674
## DO 1.005254
## ESC -1.464948
## LC -10.757089
## WC -10.034635
##
## with conditional variances for "Observer" "Plot"
## you can store the model results to objects
obs.ranef <- ranef(nm)$Observer
plot.ranef <- ranef(nm)$Plot
## the mean of these values should be close to 0
mean(obs.ranef[[1]])

## [1] 3.494283e-13
mean(plot.ranef[[1]])

## [1] -7.280149e-12
## quick model diagnostics
## extract predicted values and plot
preds <- predict(nm)
ggplot() +
  geom_histogram(mapping = aes(preds)) +
  theme_minimal()

## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.

```



```
# we want these to look normally distributed
```

```
## extract residuals and plot
```

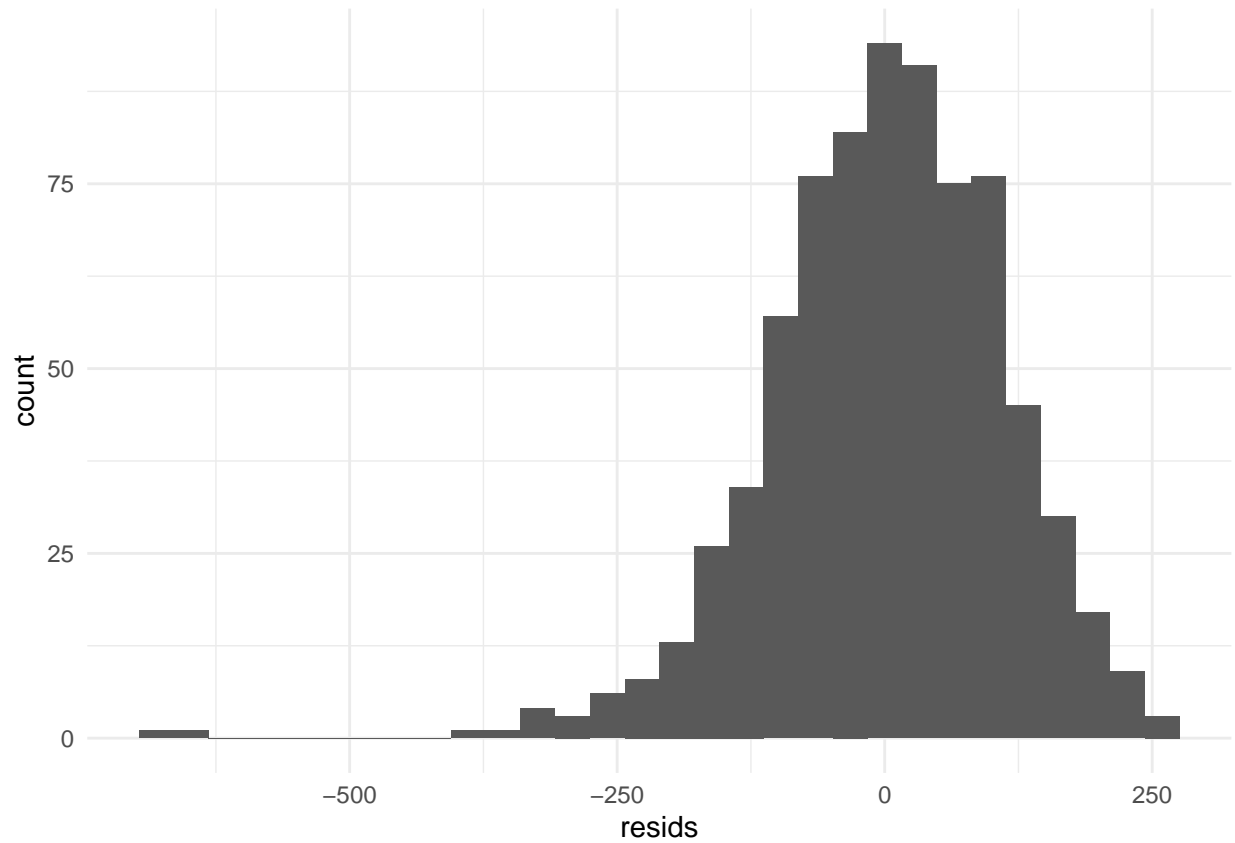
```
resids <- residuals(nm)
```

```
ggplot() +
```

```
  geom_histogram(mapping = aes(resids)) +
```

```
  theme_minimal()
```

```
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```

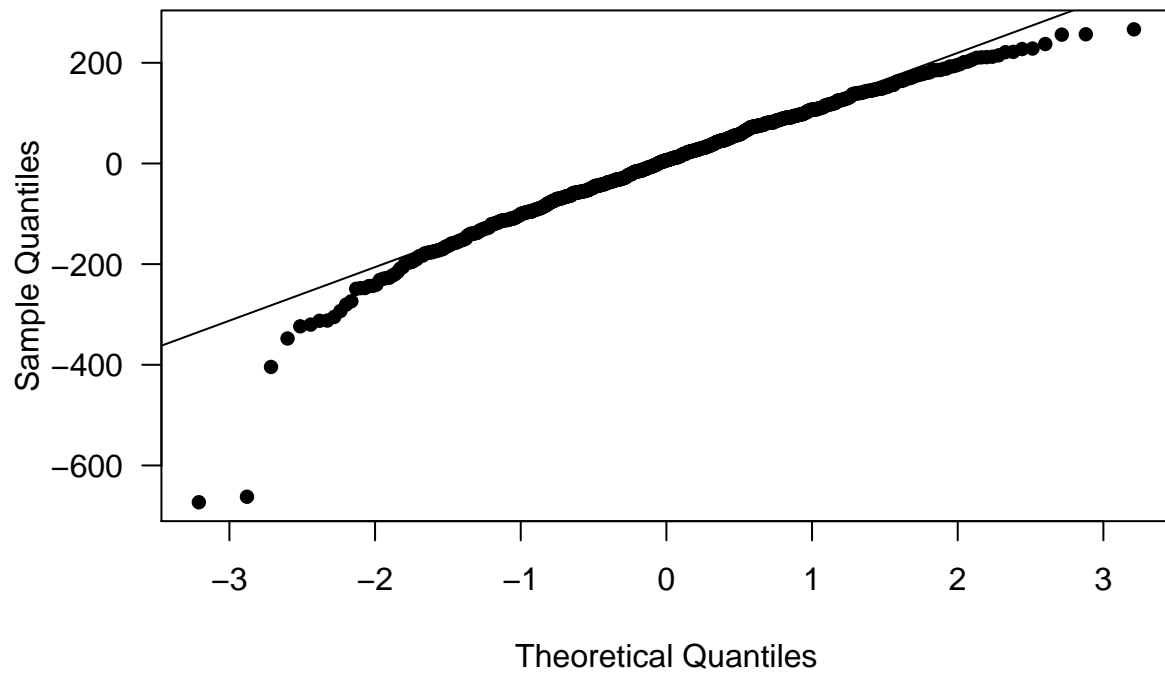


```
# we want these to look normally distributed

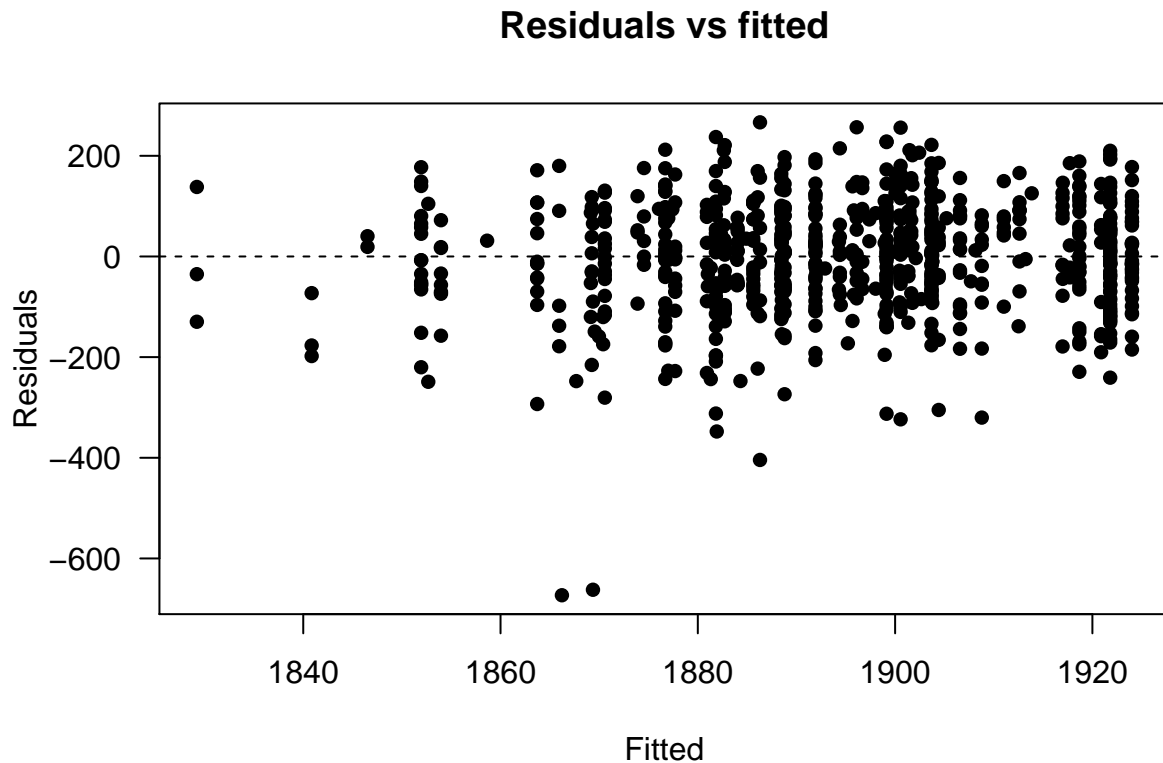
## Q-Q Plots
# ## set plot area
# par(mai = c(0.9, 0.9, 0.6, 0.1),
#     omi = c(0, 0, 0, 0),
#     mfrow = c(1,2), cex.lab = 1.2)

# qq resids
qqnorm(resids, main = "QQ plot (residuals)", las = 1, pch = 16)
qqline(resids)
```

QQ plot (residuals)



```
## plot residuals versus fitted values
yh <- fitted(nm)
plot(yh, resid, las = 1, pch = 16,
     xlab = "Fitted", ylab = "Residuals",
     main = "Residuals vs fitted")
abline(h=0, lty = "dashed")
```



Egg Order-only Model

This model includes observer and plot as random effects and egg order as a fixed effect.

```
## run the model
order.mod <- lmer(Size ~ EggOrder + (1 | Observer) + (1 | Plot), data = df)

## look at model summary and estimates
sumary(order.mod)
```

```
## Fixed Effects:
##               coef.est coef.se
## (Intercept)  1877.38   12.31
## EggOrderEgg2   31.04   10.14
##
## Random Effects:
## Groups   Name      Std.Dev.
## Observer (Intercept)  18.15
## Plot      (Intercept)  24.02
## Residual                111.29
## ---
## number of obs: 753, groups: Observer, 27; Plot, 8
## AIC = 9249.3, DIC = 9265.6
## deviance = 9252.5
```



```
coef(summary(order.mod))
```

```
##              Estimate Std. Error   t value
## (Intercept) 1877.38022   12.31228 152.480334
## EggOrderEgg2  31.03979   10.13834   3.061624
```

```
VarCorr(order.mod)
```

```
## Groups   Name      Std.Dev.
## Observer (Intercept) 18.149
## Plot      (Intercept) 24.018
## Residual                      111.286
```

```
ranef(order.mod)
```

```
## $Observer
##      (Intercept)
## AAY 16.3995440
## AJB -3.8561645
## AJD 13.2643062
## AML  1.3889281
## CAC -8.9415957
## CEH -5.3878571
## CEK -3.9207379
## CLE -5.8069434
## DLW -10.7080014
## EWW  7.0226116
## GRK -1.3811942
## JAH 20.0523198
## KMR  1.1207978
## KWB -3.7789003
## LAH -0.7457820
## MEJ -6.3933777
## MGB  2.3718992
## NAG -15.5908573
## PTL -7.8956364
## RER  8.7443717
## REW -2.6758168
## SAA  4.2279036
## SFC  1.6627498
## SJK 10.2403593
## SKT -23.7215706
## SLA  0.7036607
## SMC 13.6049837
##
## $Plot
##      (Intercept)
## APNC 21.808429
## BH   12.045434
## BT  -32.536067
## CC   19.167373
## DO   1.045153
## ESC  -1.099696
## LC  -10.595856
## WC   -9.834771
```

```
##
## with conditional variances for "Observer" "Plot"
```

Fit Candidate Models

```
## create data frame specifying predictors to include
predictors <- as.data.frame(matrix(c(FALSE, TRUE), 2, 7)) # 7 potential predictors (includes EggOrder)

## add column names
cov_names <- colnames(predictors) <- colnames(df[,5:11])

## create set of all possible combinations
full_set <- expand.grid(predictors) # 128 combinations

## select models with correlated predictors
ii <- which(full_set$ANCHL + full_set$NPGO == 2 |
            full_set$BEUTI + full_set$PDO == 2 |
            full_set$NPGO + full_set$SST == 2 |
            full_set$ONI + full_set$PDO == 2 |
            full_set$ONI + full_set$SST == 2 |
            full_set$PDO + full_set$SST == 2 ) # 90 models

## create reduced set of models and convert to a matrix for easier indexing
use_set <- as.matrix(full_set[-ii,]) # 38 models

## number of models in set
(n_mods <- nrow(use_set)) # 38 models out of potential 64

## [1] 38

## create empty matrix for storing results
mod_res <- matrix(NA, n_mods, 2)
colnames(mod_res) <- c("AIC", "BIC")

## fit models and store AIC & BIC
for(i in 1:n_mods) {
  if(i == 1) {
    fmla <- "Size ~ 1 + (1 | Observer) + (1 | Plot)"
  } else {
    fmla <- paste("Size ~ (1 | Observer) + (1 | Plot) +", paste(cov_names[use_set[i,]], collapse = " +
  )
  mod_fit <- lmer(as.formula(fmla), data = df)
  mod_res[i,"AIC"] <- AIC(mod_fit)
  mod_res[i,"BIC"] <- BIC(mod_fit)
}

## create empty matrix for storing results
delta_res <- matrix(NA, n_mods, 2)
colnames(delta_res) <- c("deltaAIC", "deltaBIC")

## convert IC to deltaIC
delta_res[, "deltaAIC"] <- mod_res[, "AIC"] - min(mod_res[, "AIC"])
delta_res[, "deltaBIC"] <- mod_res[, "BIC"] - min(mod_res[, "BIC"])
(delta_res <- round(delta_res, 2)) # round results
```

```
##           deltaAIC deltaBIC
## [1,]      34.73    17.22
## [2,]      20.95     8.07
## [3,]      26.11    13.22
## [4,]      12.95     4.68
## [5,]      31.68    18.79
## [6,]      17.88     9.62
## [7,]      17.06     8.80
## [8,]       4.26     0.62
## [9,]      23.16    10.27
## [10,]     9.80     1.54
## [11,]     16.92     8.65
## [12,]      3.64     0.00
## [13,]     30.68    17.79
## [14,]     16.54     8.28
## [15,]     19.31    11.04
## [16,]      5.78     2.13
## [17,]     27.17    18.90
## [18,]     12.96     9.32
## [19,]     13.18     9.54
## [20,]      0.10     1.08
## [21,]     19.84    11.57
## [22,]      6.26     2.62
## [23,]     13.10     9.45
## [24,]      0.00     0.98
## [25,]     31.49    18.60
## [26,]     17.68     9.41
## [27,]     19.83    11.57
## [28,]      6.54     2.90
## [29,]     19.66    11.39
## [30,]      6.19     2.55
## [31,]     30.95    18.06
## [32,]     17.35     9.09
## [33,]     21.80    13.54
## [34,]      8.55     4.91
## [35,]     27.00    18.74
## [36,]     13.50     9.86
## [37,]     13.23     9.59
## [38,]      0.37     1.36
```

```
## create df with mod results
```

```
mp <- as.data.frame(use_set)
```

```
for (i in 1:length(mp)) {
  mp[[i]] <- str_replace(mp[[i]], "TRUE", colnames(mp)[i])
}
```

```
for (i in 1:length(mp)) {
  mp[[i]] <- str_replace(mp[[i]], "FALSE", " ")
}
```

```
mpfe <- mp %>%
```

```
  mutate(FEs = paste(EggOrder, ANCHL, BEUTI, NPGO, ONI, PDO, SST, sep = " "))
```

```

usmr <- mpfe %>%
  mutate(k = as.vector(rowSums(use_set) + 2)) %>% #2 check this
  mutate(modelno = 1:38)

allm <- as.data.frame(delta_res) %>%
  mutate(modelno = 1:38) %>%
  arrange(deltaAIC) %>%
  left_join(usmr, by = "modelno") %>%
  dplyr::select(modelno, FEs, k, deltaAIC, deltaBIC)

## create df with top models
bestm <- allm %>%
  filter(deltaAIC <= 2)

```

There are three competitive models. All three models include EggOrder and BEUTI as a predictor.

Top Models

Here we look at the top three competitive models.

First Top Model

```

##run model
bm1 <- lmer(Size ~ EggOrder + BEUTI + NPGO + ONI + (1 | Observer) + (1 | Plot), data = df)
## look at model output and estimates
summary(bm1)

## Linear mixed model fit by REML ['lmerMod']
## Formula: Size ~ EggOrder + BEUTI + NPGO + ONI + (1 | Observer) + (1 |
##      Plot)
##      Data: df
##
## REML criterion at convergence: 9212.3
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -6.2281 -0.5687  0.0450  0.6626  2.3189
##
## Random effects:
##      Groups      Name              Variance Std.Dev.
##      Observer (Intercept)    132.2     11.50
##      Plot      (Intercept)    349.9     18.71
##      Residual                12365.8    111.20
## Number of obs: 753, groups:  Observer, 27; Plot, 8
##
## Fixed effects:
##              Estimate Std. Error t value
## (Intercept)  1889.549     10.362  182.358
## EggOrderEgg2    29.910     10.123   2.955
## BEUTI          11.104      6.859   1.619
## NPGO          -22.759      6.732  -3.381
## ONI             2.972      5.913   0.503
##

```

```

## Correlation of Fixed Effects:
##      (Intr) EggOE2 BEUTI  NPGO
## EggOrderEgg2 -0.218
## BEUTI        0.146 -0.042
## NPGO         -0.252  0.031 -0.597
## ONI          0.109 -0.052  0.700 -0.457

coef(bm1)

## $Observer
##      (Intercept) EggOrderEgg2  BEUTI      NPGO      ONI
## AAY      1898.226      29.90976 11.1043 -22.75851 2.971758
## AJB      1887.490      29.90976 11.1043 -22.75851 2.971758
## AJD      1893.335      29.90976 11.1043 -22.75851 2.971758
## AML      1889.849      29.90976 11.1043 -22.75851 2.971758
## CAC      1882.309      29.90976 11.1043 -22.75851 2.971758
## CEH      1886.834      29.90976 11.1043 -22.75851 2.971758
## CEK      1888.526      29.90976 11.1043 -22.75851 2.971758
## CLE      1885.842      29.90976 11.1043 -22.75851 2.971758
## DLW      1886.151      29.90976 11.1043 -22.75851 2.971758
## EWW      1890.804      29.90976 11.1043 -22.75851 2.971758
## GRK      1888.900      29.90976 11.1043 -22.75851 2.971758
## JAH      1894.719      29.90976 11.1043 -22.75851 2.971758
## KMR      1890.363      29.90976 11.1043 -22.75851 2.971758
## KWB      1893.889      29.90976 11.1043 -22.75851 2.971758
## LAH      1889.444      29.90976 11.1043 -22.75851 2.971758
## MEJ      1887.006      29.90976 11.1043 -22.75851 2.971758
## MGB      1890.467      29.90976 11.1043 -22.75851 2.971758
## NAG      1885.550      29.90976 11.1043 -22.75851 2.971758
## PTL      1885.434      29.90976 11.1043 -22.75851 2.971758
## RER      1891.692      29.90976 11.1043 -22.75851 2.971758
## REW      1887.998      29.90976 11.1043 -22.75851 2.971758
## SAA      1900.548      29.90976 11.1043 -22.75851 2.971758
## SFC      1889.011      29.90976 11.1043 -22.75851 2.971758
## SJK      1893.045      29.90976 11.1043 -22.75851 2.971758
## SKT      1879.078      29.90976 11.1043 -22.75851 2.971758
## SLA      1890.326      29.90976 11.1043 -22.75851 2.971758
## SMC      1890.980      29.90976 11.1043 -22.75851 2.971758
##
## $Plot
##      (Intercept) EggOrderEgg2  BEUTI      NPGO      ONI
## APNC      1906.937      29.90976 11.1043 -22.75851 2.971758
## BH         1899.167      29.90976 11.1043 -22.75851 2.971758
## BT         1870.282      29.90976 11.1043 -22.75851 2.971758
## CC         1904.589      29.90976 11.1043 -22.75851 2.971758
## DO         1884.902      29.90976 11.1043 -22.75851 2.971758
## ESC        1888.654      29.90976 11.1043 -22.75851 2.971758
## LC         1877.422      29.90976 11.1043 -22.75851 2.971758
## WC         1884.437      29.90976 11.1043 -22.75851 2.971758
##
## attr(,"class")
## [1] "coef.mer"

ranef(bm1)

## $Observer

```

```

##      (Intercept)
## AAY      8.6772576
## AJB     -2.0586077
## AJD      3.7860407
## AML      0.3004975
## CAC     -7.2399067
## CEH     -2.7152268
## CEK     -1.0229503
## CLE     -3.7068839
## DLW     -3.3975244
## EWW      1.2556806
## GRK     -0.6489871
## JAH      5.1701353
## KMR      0.8141630
## KWB      4.3404043
## LAH     -0.1049257
## MEJ     -2.5423008
## MGB      0.9181979
## NAG     -3.9985573
## PTL     -4.1146182
## RER      2.1428225
## REW     -1.5503674
## SAA     10.9992085
## SFC     -0.5374726
## SJK      3.4960544
## SKT    -10.4707426
## SLA      0.7769067
## SMC      1.4317026
##
## $Plot
##      (Intercept)
## APNC     17.3883993
## BH        9.6185095
## BT     -19.2669360
## CC       15.0399622
## DO       -4.6469222
## ESC      -0.8947288
## LC      -12.1267777
## WC       -5.1115063
##
## with conditional variances for "Observer" "Plot"
model.matrix(bm1)
##      (Intercept) EggOrderEgg2      BEUTI      NPGO      ONI
## 1              1            0 0.6675426 0.3555739 -0.61651548
## 2              1            0 0.6675426 0.3555739 -0.61651548
## 3              1            0 0.6675426 0.3555739 -0.61651548
## 4              1            0 -0.1856417 1.5217365 0.71011823
## 5              1            0 -0.1856417 1.5217365 0.71011823
## 6              1            0 -0.1856417 1.5217365 0.71011823
## 7              1            0 -0.1856417 1.5217365 0.71011823
## 8              1            0 -0.1856417 1.5217365 0.71011823
## 9              1            0 -0.1856417 1.5217365 0.71011823
## 10             1            0 -0.1856417 1.5217365 0.71011823

```

[illegible]

## 65	1	0	-0.1856417	1.5217365	0.71011823
## 66	1	0	-0.1856417	1.5217365	0.71011823
## 67	1	0	-0.1856417	1.5217365	0.71011823
## 68	1	0	-0.1856417	1.5217365	0.71011823
## 69	1	0	-0.1856417	1.5217365	0.71011823
## 70	1	0	-0.1856417	1.5217365	0.71011823
## 71	1	0	-0.1856417	1.5217365	0.71011823
## 72	1	0	-0.1856417	1.5217365	0.71011823
## 73	1	0	-0.1856417	1.5217365	0.71011823
## 74	1	0	-0.1856417	1.5217365	0.71011823
## 75	1	0	-0.1856417	1.5217365	0.71011823
## 76	1	0	-0.1856417	1.5217365	0.71011823
## 77	1	0	-0.1856417	1.5217365	0.71011823
## 78	1	0	-0.1856417	1.5217365	0.71011823
## 79	1	0	-0.1856417	1.5217365	0.71011823
## 80	1	0	-0.1856417	1.5217365	0.71011823
## 81	1	0	-0.1856417	1.5217365	0.71011823
## 82	1	0	-0.1856417	1.5217365	0.71011823
## 83	1	0	-0.1856417	1.5217365	0.71011823
## 84	1	0	-0.1856417	1.5217365	0.71011823
## 85	1	0	-0.1856417	1.5217365	0.71011823
## 86	1	0	-0.1856417	1.5217365	0.71011823
## 87	1	0	-0.1856417	1.5217365	0.71011823
## 88	1	0	-0.1856417	1.5217365	0.71011823
## 89	1	0	-0.1856417	1.5217365	0.71011823
## 90	1	0	-0.1856417	1.5217365	0.71011823
## 91	1	0	-0.1856417	1.5217365	0.71011823
## 92	1	0	-0.1856417	1.5217365	0.71011823
## 93	1	0	-0.1856417	1.5217365	0.71011823
## 94	1	0	-0.1856417	1.5217365	0.71011823
## 95	1	0	-0.1856417	1.5217365	0.71011823
## 96	1	0	-0.1856417	1.5217365	0.71011823
## 97	1	0	-0.1856417	1.5217365	0.71011823
## 98	1	0	-0.1856417	1.5217365	0.71011823
## 99	1	0	-0.1856417	1.5217365	0.71011823
## 100	1	0	-0.1856417	1.5217365	0.71011823
## 101	1	0	-0.1856417	1.5217365	0.71011823
## 102	1	0	-0.1856417	1.5217365	0.71011823
## 103	1	0	-0.1856417	1.5217365	0.71011823
## 104	1	0	-0.1856417	1.5217365	0.71011823
## 105	1	0	-0.1856417	1.5217365	0.71011823
## 106	1	0	-0.1856417	1.5217365	0.71011823
## 107	1	0	-0.1856417	1.5217365	0.71011823
## 108	1	0	-0.1856417	1.5217365	0.71011823
## 109	1	0	-0.1856417	1.5217365	0.71011823
## 110	1	0	-0.1856417	1.5217365	0.71011823
## 111	1	0	-0.1856417	1.5217365	0.71011823
## 112	1	0	-0.1856417	1.5217365	0.71011823
## 113	1	0	-0.1902758	0.4783374	-1.35024026
## 114	1	0	-0.1902758	0.4783374	-1.35024026
## 115	1	0	-0.1902758	0.4783374	-1.35024026
## 116	1	0	-0.1902758	0.4783374	-1.35024026
## 117	1	0	-0.1902758	0.4783374	-1.35024026
## 118	1	0	-0.1902758	0.4783374	-1.35024026

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

## 701	1	1	-0.1856417	1.5217365	0.71011823
## 702	1	1	-0.1856417	1.5217365	0.71011823
## 703	1	1	-0.1856417	1.5217365	0.71011823
## 704	1	1	-0.1856417	1.5217365	0.71011823
## 705	1	1	-0.1856417	1.5217365	0.71011823
## 706	1	1	-0.1856417	1.5217365	0.71011823
## 707	1	1	-0.1856417	1.5217365	0.71011823
## 708	1	1	-0.1856417	1.5217365	0.71011823
## 709	1	1	-0.1856417	1.5217365	0.71011823
## 710	1	1	-0.1856417	1.5217365	0.71011823
## 711	1	1	-0.1856417	1.5217365	0.71011823
## 712	1	1	-0.1856417	1.5217365	0.71011823
## 713	1	1	-0.1856417	1.5217365	0.71011823
## 714	1	1	-0.1856417	1.5217365	0.71011823
## 715	1	1	-0.1856417	1.5217365	0.71011823
## 716	1	1	-0.1856417	1.5217365	0.71011823
## 717	1	1	-0.1856417	1.5217365	0.71011823
## 718	1	1	-0.1856417	1.5217365	0.71011823
## 719	1	1	-0.1856417	1.5217365	0.71011823
## 720	1	1	-0.1856417	1.5217365	0.71011823
## 721	1	1	-0.1856417	1.5217365	0.71011823
## 722	1	1	-0.1856417	1.5217365	0.71011823
## 723	1	1	-0.1856417	1.5217365	0.71011823
## 724	1	1	-0.1856417	1.5217365	0.71011823
## 725	1	1	-0.1856417	1.5217365	0.71011823
## 726	1	1	-0.1856417	1.5217365	0.71011823
## 727	1	1	-0.1902758	0.4783374	-1.35024026
## 728	1	1	-0.1902758	0.4783374	-1.35024026
## 729	1	1	-0.1902758	0.4783374	-1.35024026
## 730	1	1	-0.1902758	0.4783374	-1.35024026
## 731	1	1	-0.1902758	0.4783374	-1.35024026
## 732	1	1	-0.1902758	0.4783374	-1.35024026
## 733	1	1	-0.1902758	0.4783374	-1.35024026
## 734	1	1	-0.1902758	0.4783374	-1.35024026
## 735	1	1	-0.1902758	0.4783374	-1.35024026
## 736	1	1	-0.1902758	0.4783374	-1.35024026
## 737	1	1	-0.1902758	0.4783374	-1.35024026
## 738	1	1	-0.1902758	0.4783374	-1.35024026
## 739	1	1	-0.1902758	0.4783374	-1.35024026
## 740	1	1	-0.1902758	0.4783374	-1.35024026
## 741	1	1	-0.1902758	0.4783374	-1.35024026
## 742	1	1	0.4290217	0.5034663	-0.77215407
## 743	1	1	0.4290217	0.5034663	-0.77215407
## 744	1	1	0.4290217	0.5034663	-0.77215407
## 745	1	1	0.4290217	0.5034663	-0.77215407
## 746	1	1	0.4290217	0.5034663	-0.77215407
## 747	1	1	0.4290217	0.5034663	-0.77215407
## 748	1	1	0.4290217	0.5034663	-0.77215407
## 749	1	1	0.4290217	0.5034663	-0.77215407
## 750	1	1	0.4290217	0.5034663	-0.77215407
## 751	1	1	0.4290217	0.5034663	-0.77215407
## 752	1	1	0.4290217	0.5034663	-0.77215407
## 753	1	1	0.4290217	0.5034663	-0.77215407
## 754	1	1	0.4290217	0.5034663	-0.77215407

## 755	1	1	0.4290217	0.5034663	-0.77215407
## 756	1	1	0.4290217	0.5034663	-0.77215407
## 757	1	1	0.4290217	0.5034663	-0.77215407
## 758	1	1	0.4290217	0.5034663	-0.77215407
## 759	1	1	0.4290217	0.5034663	-0.77215407
## 760	1	1	0.4290217	0.5034663	-0.77215407
## 761	1	1	0.4290217	0.5034663	-0.77215407
## 762	1	1	0.4290217	0.5034663	-0.77215407
## 763	1	1	0.4290217	0.5034663	-0.77215407
## 764	1	1	0.4290217	0.5034663	-0.77215407
## 765	1	1	0.4290217	0.5034663	-0.77215407
## 766	1	1	0.4290217	0.5034663	-0.77215407
## 767	1	1	0.4290217	0.5034663	-0.77215407
## 768	1	1	2.1517850	0.8723367	-0.60663366
## 769	1	1	2.1517850	0.8723367	-0.60663366
## 770	1	1	2.1517850	0.8723367	-0.60663366
## 771	1	1	2.1517850	0.8723367	-0.60663366
## 772	1	1	2.1517850	0.8723367	-0.60663366
## 773	1	1	2.1517850	0.8723367	-0.60663366
## 774	1	1	2.1517850	0.8723367	-0.60663366
## 775	1	1	2.1517850	0.8723367	-0.60663366
## 776	1	1	2.1517850	0.8723367	-0.60663366
## 777	1	1	2.1517850	0.8723367	-0.60663366
## 778	1	1	2.1517850	0.8723367	-0.60663366
## 779	1	1	2.1517850	0.8723367	-0.60663366
## 780	1	1	2.1517850	0.8723367	-0.60663366
## 781	1	1	2.1517850	0.8723367	-0.60663366
## 782	1	1	2.1517850	0.8723367	-0.60663366
## 783	1	1	2.1517850	0.8723367	-0.60663366
## 784	1	1	2.1517850	0.8723367	-0.60663366
## 785	1	1	2.1517850	0.8723367	-0.60663366
## 786	1	1	2.1517850	0.8723367	-0.60663366
## 787	1	1	2.1517850	0.8723367	-0.60663366
## 788	1	1	2.1517850	0.8723367	-0.60663366
## 789	1	1	-0.3861161	-1.0216977	-0.22865423
## 790	1	1	-0.3861161	-1.0216977	-0.22865423
## 791	1	1	-0.3861161	-1.0216977	-0.22865423
## 792	1	1	-0.3861161	-1.0216977	-0.22865423
## 793	1	1	-0.3861161	-1.0216977	-0.22865423
## 794	1	1	-0.3861161	-1.0216977	-0.22865423
## 795	1	1	-0.3861161	-1.0216977	-0.22865423
## 796	1	1	-0.3861161	-1.0216977	-0.22865423
## 797	1	1	-0.3861161	-1.0216977	-0.22865423
## 798	1	1	-0.3861161	-1.0216977	-0.22865423
## 799	1	1	-0.3861161	-1.0216977	-0.22865423
## 800	1	1	-0.3861161	-1.0216977	-0.22865423
## 801	1	1	-0.3861161	-1.0216977	-0.22865423
## 802	1	1	-0.3861161	-1.0216977	-0.22865423
## 803	1	1	-0.3861161	-1.0216977	-0.22865423
## 804	1	1	-0.3861161	-1.0216977	-0.22865423
## 805	1	1	-0.3861161	-1.0216977	-0.22865423
## 831	1	1	-1.3133684	-0.1631466	1.79464745
## 832	1	1	-1.3133684	-0.1631466	1.79464745
## 833	1	1	-1.3133684	-0.1631466	1.79464745

```

## 834      1      1 -1.3133684 -0.1631466 1.79464745
## 835      1      1 -1.3133684 -0.1631466 1.79464745
## 836      1      1 -1.3133684 -0.1631466 1.79464745
## 837      1      1 -1.3133684 -0.1631466 1.79464745
## 838      1      1 -1.3133684 -0.1631466 1.79464745
## 839      1      1 -1.3133684 -0.1631466 1.79464745
## 840      1      1 -1.3133684 -0.1631466 1.79464745
## 841      1      1 -1.3133684 -0.1631466 1.79464745
## 842      1      1 -1.3133684 -0.1631466 1.79464745
## 843      1      1 -1.3133684 -0.1631466 1.79464745
## 844      1      1 -1.3133684 -0.1631466 1.79464745
## 845      1      1 -1.3133684 -0.1631466 1.79464745
## 846      1      1 -1.3133684 -0.1631466 1.79464745
## 847      1      1 -1.3133684 -0.1631466 1.79464745
## 848      1      1 -1.3133684 -0.1631466 1.79464745
## 849      1      1 -1.3133684 -0.1631466 1.79464745
## 850      1      1 -1.3133684 -0.1631466 1.79464745
## 851      1      1 -1.3133684 -0.1631466 1.79464745
## 852      1      1 -1.3133684 -0.1631466 1.79464745
## 853      1      1 -1.3133684 -0.1631466 1.79464745
## 854      1      1 -1.3133684 -0.1631466 1.79464745
## 855      1      1 -1.3133684 -0.1631466 1.79464745
## 856      1      1 -1.3133684 -0.1631466 1.79464745
## 857      1      1 -1.3133684 -0.1631466 1.79464745
## 858      1      1 -1.3133684 -0.1631466 1.79464745
## 859      1      1 -1.3133684 -0.1631466 1.79464745
## 860      1      1 -1.3133684 -0.1631466 1.79464745
## 861      1      1 -0.8183743 -1.2455211 0.04309569
## 862      1      1 -0.8183743 -1.2455211 0.04309569
## 863      1      1 -0.8183743 -1.2455211 0.04309569
## 864      1      1 -0.8183743 -1.2455211 0.04309569
## 865      1      1 -0.8183743 -1.2455211 0.04309569
## 866      1      1 -0.8183743 -1.2455211 0.04309569
## 867      1      1 -0.8183743 -1.2455211 0.04309569
## 868      1      1 -0.8183743 -1.2455211 0.04309569
## 869      1      1 -0.8183743 -1.2455211 0.04309569
## 870      1      1 -0.8183743 -1.2455211 0.04309569
## 871      1      1 -0.8183743 -1.2455211 0.04309569
## 872      1      1 -0.8183743 -1.2455211 0.04309569
## 873      1      1 -0.8183743 -1.2455211 0.04309569
## 874      1      1 -0.8183743 -1.2455211 0.04309569
## attr("assign")
## [1] 0 1 2 3 4
## attr("contrasts")
## attr("contrasts")$EggOrder
## [1] "contr.treatment"
##
## attr("msgScaleX")
## character(0)

```

Second Top Model

```

##run model
bm2 <- lmer(Size ~ EggOrder + ANCHL + BEUTI + ONI + (1 | Observer) + (1 | Plot), data = df)

```

```
## look at model output and estimates
```

```
summary(bm2)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: Size ~ EggOrder + ANCHL + BEUTI + ONI + (1 | Observer) + (1 |
##   Plot)
##   Data: df
##
## REML criterion at convergence: 9212.4
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -6.2131 -0.5812  0.0618  0.6658  2.3488
##
## Random effects:
##   Groups   Name                Variance Std.Dev.
##   Observer (Intercept)    109.3     10.45
##   Plot      (Intercept)    337.1     18.36
##   Residual                  12380.3   111.27
## Number of obs: 753, groups:  Observer, 27; Plot, 8
##
## Fixed effects:
##              Estimate Std. Error t value
## (Intercept)  1884.733      9.882 190.716
## EggOrderEgg2   29.894     10.124   2.953
## ANCHL          22.170      6.378   3.476
## BEUTI          9.546      6.514   1.465
## ONI           -5.347      5.244  -1.020
##
## Correlation of Fixed Effects:
##              (Intr) EggOE2 ANCHL  BEUTI
## EggOrdrEgg2 -0.223
## ANCHL        0.104 -0.026
## BEUTI        0.056 -0.040  0.544
## ONI         -0.004 -0.044  0.033  0.521
```

```
coef(bm2)
```

```
## $Observer
##      (Intercept) EggOrderEgg2   ANCHL   BEUTI   ONI
## AAY      1893.437      29.8937 22.17002 9.546142 -5.346525
## AJB      1883.313      29.8937 22.17002 9.546142 -5.346525
## AJD      1887.079      29.8937 22.17002 9.546142 -5.346525
## AML      1885.024      29.8937 22.17002 9.546142 -5.346525
## CAC      1879.133      29.8937 22.17002 9.546142 -5.346525
## CEH      1882.567      29.8937 22.17002 9.546142 -5.346525
## CEK      1883.783      29.8937 22.17002 9.546142 -5.346525
## CLE      1881.542      29.8937 22.17002 9.546142 -5.346525
## DLW      1881.590      29.8937 22.17002 9.546142 -5.346525
## EWW      1886.065      29.8937 22.17002 9.546142 -5.346525
## GRK      1884.338      29.8937 22.17002 9.546142 -5.346525
## JAH      1888.808      29.8937 22.17002 9.546142 -5.346525
## KMR      1886.197      29.8937 22.17002 9.546142 -5.346525
## KWB      1887.317      29.8937 22.17002 9.546142 -5.346525
```

```
## LAH      1884.619      29.8937 22.17002 9.546142 -5.346525
## MEJ      1882.600      29.8937 22.17002 9.546142 -5.346525
## MGB      1885.551      29.8937 22.17002 9.546142 -5.346525
## NAG      1880.707      29.8937 22.17002 9.546142 -5.346525
## PTL      1881.487      29.8937 22.17002 9.546142 -5.346525
## RER      1886.401      29.8937 22.17002 9.546142 -5.346525
## REW      1884.001      29.8937 22.17002 9.546142 -5.346525
## SAA      1893.001      29.8937 22.17002 9.546142 -5.346525
## SFC      1884.442      29.8937 22.17002 9.546142 -5.346525
## SJK      1887.579      29.8937 22.17002 9.546142 -5.346525
## SKT      1875.007      29.8937 22.17002 9.546142 -5.346525
## SLA      1885.729      29.8937 22.17002 9.546142 -5.346525
## SMC      1886.473      29.8937 22.17002 9.546142 -5.346525
```

```
##
```

```
## $Plot
```

```
##      (Intercept) EggOrderEgg2      ANCHL      BEUTI      ONI
## APNC      1901.362      29.8937 22.17002 9.546142 -5.346525
## BH        1894.141      29.8937 22.17002 9.546142 -5.346525
## BT        1864.569      29.8937 22.17002 9.546142 -5.346525
## CC        1899.300      29.8937 22.17002 9.546142 -5.346525
## DO        1880.738      29.8937 22.17002 9.546142 -5.346525
## ESC       1883.886      29.8937 22.17002 9.546142 -5.346525
## LC        1874.590      29.8937 22.17002 9.546142 -5.346525
## WC        1879.278      29.8937 22.17002 9.546142 -5.346525
```

```
##
```

```
## attr("class")
```

```
## [1] "coef.mer"
```

```
ranef(bm2)
```

```
## $Observer
```

```
##      (Intercept)
```

```
## AAY      8.7037184
## AJB     -1.4199693
## AJD      2.3463483
## AML      0.2909773
## CAC     -5.6002721
## CEH     -2.1655779
## CEK     -0.9502841
## CLE     -3.1904763
## DLW     -3.1426862
## EWW      1.3322116
## GRK     -0.3949463
## JAH      4.0746183
## KMR      1.4645020
## KWB      2.5836403
## LAH     -0.1143350
## MEJ     -2.1334064
## MGB      0.8175811
## NAG     -4.0257174
## PTL     -3.2458735
## RER      1.6680647
## REW     -0.7319769
## SAA      8.2680905
## SFC     -0.2911164
```

```
## SJK    2.8460718
## SKT   -9.7256778
## SLA    0.9959990
## SMC    1.7404923
##
## $Plot
##      (Intercept)
## APNC   16.6291061
## BH      9.4084040
## BT   -20.1642390
## CC     14.5669458
## DO     -3.9950551
## ESC    -0.8469532
## LC    -10.1427500
## WC     -5.4554586
##
## with conditional variances for "Observer" "Plot"
```

```
model.matrix(bm2)
```

##	(Intercept)	EggOrderEgg2	ANCHL	BEUTI	ONI
## 1	1	0	-0.4071777	0.6675426	-0.61651548
## 2	1	0	-0.4071777	0.6675426	-0.61651548
## 3	1	0	-0.4071777	0.6675426	-0.61651548
## 4	1	0	-0.8196345	-0.1856417	0.71011823
## 5	1	0	-0.8196345	-0.1856417	0.71011823
## 6	1	0	-0.8196345	-0.1856417	0.71011823
## 7	1	0	-0.8196345	-0.1856417	0.71011823
## 8	1	0	-0.8196345	-0.1856417	0.71011823
## 9	1	0	-0.8196345	-0.1856417	0.71011823
## 10	1	0	-0.8196345	-0.1856417	0.71011823
## 11	1	0	-0.8196345	-0.1856417	0.71011823
## 12	1	0	-0.8196345	-0.1856417	0.71011823
## 13	1	0	-0.8196345	-0.1856417	0.71011823
## 14	1	0	-0.8196345	-0.1856417	0.71011823
## 15	1	0	-0.8196345	-0.1856417	0.71011823
## 16	1	0	-0.8196345	-0.1856417	0.71011823
## 17	1	0	-0.8196345	-0.1856417	0.71011823
## 18	1	0	-0.8196345	-0.1856417	0.71011823
## 19	1	0	-0.8196345	-0.1856417	0.71011823
## 20	1	0	-0.8196345	-0.1856417	0.71011823
## 21	1	0	-0.8196345	-0.1856417	0.71011823
## 22	1	0	-0.8196345	-0.1856417	0.71011823
## 23	1	0	-0.8196345	-0.1856417	0.71011823
## 24	1	0	-0.8196345	-0.1856417	0.71011823
## 25	1	0	-0.8196345	-0.1856417	0.71011823
## 26	1	0	-0.8196345	-0.1856417	0.71011823
## 27	1	0	-0.8196345	-0.1856417	0.71011823
## 28	1	0	-0.8196345	-0.1856417	0.71011823
## 29	1	0	-0.8196345	-0.1856417	0.71011823
## 30	1	0	-0.8196345	-0.1856417	0.71011823
## 31	1	0	-0.8196345	-0.1856417	0.71011823
## 32	1	0	-0.8196345	-0.1856417	0.71011823
## 33	1	0	-0.8196345	-0.1856417	0.71011823
## 34	1	0	-0.8196345	-0.1856417	0.71011823

[illegible]

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[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

## 671	1	0	1.9637172	-0.8183743	0.04309569
## 672	1	0	1.9637172	-0.8183743	0.04309569
## 673	1	0	1.9637172	-0.8183743	0.04309569
## 674	1	0	1.9637172	-0.8183743	0.04309569
## 675	1	0	1.9637172	-0.8183743	0.04309569
## 676	1	0	1.9637172	-0.8183743	0.04309569
## 677	1	0	1.9637172	-0.8183743	0.04309569
## 678	1	0	1.9637172	-0.8183743	0.04309569
## 679	1	0	1.9637172	-0.8183743	0.04309569
## 680	1	0	1.9637172	-0.8183743	0.04309569
## 681	1	0	1.9637172	-0.8183743	0.04309569
## 682	1	0	1.9637172	-0.8183743	0.04309569
## 683	1	0	1.9637172	-0.8183743	0.04309569
## 684	1	0	1.9637172	-0.8183743	0.04309569
## 685	1	0	1.9637172	-0.8183743	0.04309569
## 686	1	0	1.9637172	-0.8183743	0.04309569
## 687	1	0	1.9637172	-0.8183743	0.04309569
## 688	1	0	1.9637172	-0.8183743	0.04309569
## 689	1	0	1.9637172	-0.8183743	0.04309569
## 690	1	0	1.9637172	-0.8183743	0.04309569
## 691	1	0	1.9637172	-0.8183743	0.04309569
## 692	1	0	1.9637172	-0.8183743	0.04309569
## 693	1	0	1.9637172	-0.8183743	0.04309569
## 694	1	0	1.9637172	-0.8183743	0.04309569
## 695	1	0	1.9637172	-0.8183743	0.04309569
## 696	1	0	1.9637172	-0.8183743	0.04309569
## 697	1	1	-0.4071777	0.6675426	-0.61651548
## 698	1	1	-0.4071777	0.6675426	-0.61651548
## 699	1	1	-0.4071777	0.6675426	-0.61651548
## 700	1	1	-0.4071777	0.6675426	-0.61651548
## 701	1	1	-0.8196345	-0.1856417	0.71011823
## 702	1	1	-0.8196345	-0.1856417	0.71011823
## 703	1	1	-0.8196345	-0.1856417	0.71011823
## 704	1	1	-0.8196345	-0.1856417	0.71011823
## 705	1	1	-0.8196345	-0.1856417	0.71011823
## 706	1	1	-0.8196345	-0.1856417	0.71011823
## 707	1	1	-0.8196345	-0.1856417	0.71011823
## 708	1	1	-0.8196345	-0.1856417	0.71011823
## 709	1	1	-0.8196345	-0.1856417	0.71011823
## 710	1	1	-0.8196345	-0.1856417	0.71011823
## 711	1	1	-0.8196345	-0.1856417	0.71011823
## 712	1	1	-0.8196345	-0.1856417	0.71011823
## 713	1	1	-0.8196345	-0.1856417	0.71011823
## 714	1	1	-0.8196345	-0.1856417	0.71011823
## 715	1	1	-0.8196345	-0.1856417	0.71011823
## 716	1	1	-0.8196345	-0.1856417	0.71011823
## 717	1	1	-0.8196345	-0.1856417	0.71011823
## 718	1	1	-0.8196345	-0.1856417	0.71011823
## 719	1	1	-0.8196345	-0.1856417	0.71011823
## 720	1	1	-0.8196345	-0.1856417	0.71011823
## 721	1	1	-0.8196345	-0.1856417	0.71011823
## 722	1	1	-0.8196345	-0.1856417	0.71011823
## 723	1	1	-0.8196345	-0.1856417	0.71011823
## 724	1	1	-0.8196345	-0.1856417	0.71011823

## 725	1	1 -0.8196345 -0.1856417 0.71011823
## 726	1	1 -0.8196345 -0.1856417 0.71011823
## 727	1	1 -0.7266228 -0.1902758 -1.35024026
## 728	1	1 -0.7266228 -0.1902758 -1.35024026
## 729	1	1 -0.7266228 -0.1902758 -1.35024026
## 730	1	1 -0.7266228 -0.1902758 -1.35024026
## 731	1	1 -0.7266228 -0.1902758 -1.35024026
## 732	1	1 -0.7266228 -0.1902758 -1.35024026
## 733	1	1 -0.7266228 -0.1902758 -1.35024026
## 734	1	1 -0.7266228 -0.1902758 -1.35024026
## 735	1	1 -0.7266228 -0.1902758 -1.35024026
## 736	1	1 -0.7266228 -0.1902758 -1.35024026
## 737	1	1 -0.7266228 -0.1902758 -1.35024026
## 738	1	1 -0.7266228 -0.1902758 -1.35024026
## 739	1	1 -0.7266228 -0.1902758 -1.35024026
## 740	1	1 -0.7266228 -0.1902758 -1.35024026
## 741	1	1 -0.7266228 -0.1902758 -1.35024026
## 742	1	1 -0.7848166 0.4290217 -0.77215407
## 743	1	1 -0.7848166 0.4290217 -0.77215407
## 744	1	1 -0.7848166 0.4290217 -0.77215407
## 745	1	1 -0.7848166 0.4290217 -0.77215407
## 746	1	1 -0.7848166 0.4290217 -0.77215407
## 747	1	1 -0.7848166 0.4290217 -0.77215407
## 748	1	1 -0.7848166 0.4290217 -0.77215407
## 749	1	1 -0.7848166 0.4290217 -0.77215407
## 750	1	1 -0.7848166 0.4290217 -0.77215407
## 751	1	1 -0.7848166 0.4290217 -0.77215407
## 752	1	1 -0.7848166 0.4290217 -0.77215407
## 753	1	1 -0.7848166 0.4290217 -0.77215407
## 754	1	1 -0.7848166 0.4290217 -0.77215407
## 755	1	1 -0.7848166 0.4290217 -0.77215407
## 756	1	1 -0.7848166 0.4290217 -0.77215407
## 757	1	1 -0.7848166 0.4290217 -0.77215407
## 758	1	1 -0.7848166 0.4290217 -0.77215407
## 759	1	1 -0.7848166 0.4290217 -0.77215407
## 760	1	1 -0.7848166 0.4290217 -0.77215407
## 761	1	1 -0.7848166 0.4290217 -0.77215407
## 762	1	1 -0.7848166 0.4290217 -0.77215407
## 763	1	1 -0.7848166 0.4290217 -0.77215407
## 764	1	1 -0.7848166 0.4290217 -0.77215407
## 765	1	1 -0.7848166 0.4290217 -0.77215407
## 766	1	1 -0.7848166 0.4290217 -0.77215407
## 767	1	1 -0.7848166 0.4290217 -0.77215407
## 768	1	1 -0.7585699 2.1517850 -0.60663366
## 769	1	1 -0.7585699 2.1517850 -0.60663366
## 770	1	1 -0.7585699 2.1517850 -0.60663366
## 771	1	1 -0.7585699 2.1517850 -0.60663366
## 772	1	1 -0.7585699 2.1517850 -0.60663366
## 773	1	1 -0.7585699 2.1517850 -0.60663366
## 774	1	1 -0.7585699 2.1517850 -0.60663366
## 775	1	1 -0.7585699 2.1517850 -0.60663366
## 776	1	1 -0.7585699 2.1517850 -0.60663366
## 777	1	1 -0.7585699 2.1517850 -0.60663366
## 778	1	1 -0.7585699 2.1517850 -0.60663366

## 779	1	1	-0.7585699	2.1517850	-0.60663366
## 780	1	1	-0.7585699	2.1517850	-0.60663366
## 781	1	1	-0.7585699	2.1517850	-0.60663366
## 782	1	1	-0.7585699	2.1517850	-0.60663366
## 783	1	1	-0.7585699	2.1517850	-0.60663366
## 784	1	1	-0.7585699	2.1517850	-0.60663366
## 785	1	1	-0.7585699	2.1517850	-0.60663366
## 786	1	1	-0.7585699	2.1517850	-0.60663366
## 787	1	1	-0.7585699	2.1517850	-0.60663366
## 788	1	1	-0.7585699	2.1517850	-0.60663366
## 789	1	1	0.9565296	-0.3861161	-0.22865423
## 790	1	1	0.9565296	-0.3861161	-0.22865423
## 791	1	1	0.9565296	-0.3861161	-0.22865423
## 792	1	1	0.9565296	-0.3861161	-0.22865423
## 793	1	1	0.9565296	-0.3861161	-0.22865423
## 794	1	1	0.9565296	-0.3861161	-0.22865423
## 795	1	1	0.9565296	-0.3861161	-0.22865423
## 796	1	1	0.9565296	-0.3861161	-0.22865423
## 797	1	1	0.9565296	-0.3861161	-0.22865423
## 798	1	1	0.9565296	-0.3861161	-0.22865423
## 799	1	1	0.9565296	-0.3861161	-0.22865423
## 800	1	1	0.9565296	-0.3861161	-0.22865423
## 801	1	1	0.9565296	-0.3861161	-0.22865423
## 802	1	1	0.9565296	-0.3861161	-0.22865423
## 803	1	1	0.9565296	-0.3861161	-0.22865423
## 804	1	1	0.9565296	-0.3861161	-0.22865423
## 805	1	1	0.9565296	-0.3861161	-0.22865423
## 831	1	1	0.7911322	-1.3133684	1.79464745
## 832	1	1	0.7911322	-1.3133684	1.79464745
## 833	1	1	0.7911322	-1.3133684	1.79464745
## 834	1	1	0.7911322	-1.3133684	1.79464745
## 835	1	1	0.7911322	-1.3133684	1.79464745
## 836	1	1	0.7911322	-1.3133684	1.79464745
## 837	1	1	0.7911322	-1.3133684	1.79464745
## 838	1	1	0.7911322	-1.3133684	1.79464745
## 839	1	1	0.7911322	-1.3133684	1.79464745
## 840	1	1	0.7911322	-1.3133684	1.79464745
## 841	1	1	0.7911322	-1.3133684	1.79464745
## 842	1	1	0.7911322	-1.3133684	1.79464745
## 843	1	1	0.7911322	-1.3133684	1.79464745
## 844	1	1	0.7911322	-1.3133684	1.79464745
## 845	1	1	0.7911322	-1.3133684	1.79464745
## 846	1	1	0.7911322	-1.3133684	1.79464745
## 847	1	1	0.7911322	-1.3133684	1.79464745
## 848	1	1	0.7911322	-1.3133684	1.79464745
## 849	1	1	0.7911322	-1.3133684	1.79464745
## 850	1	1	0.7911322	-1.3133684	1.79464745
## 851	1	1	0.7911322	-1.3133684	1.79464745
## 852	1	1	0.7911322	-1.3133684	1.79464745
## 853	1	1	0.7911322	-1.3133684	1.79464745
## 854	1	1	0.7911322	-1.3133684	1.79464745
## 855	1	1	0.7911322	-1.3133684	1.79464745
## 856	1	1	0.7911322	-1.3133684	1.79464745
## 857	1	1	0.7911322	-1.3133684	1.79464745

```
## 858      1      1 0.7911322 -1.3133684 1.79464745
## 859      1      1 0.7911322 -1.3133684 1.79464745
## 860      1      1 0.7911322 -1.3133684 1.79464745
## 861      1      1 1.9637172 -0.8183743 0.04309569
## 862      1      1 1.9637172 -0.8183743 0.04309569
## 863      1      1 1.9637172 -0.8183743 0.04309569
## 864      1      1 1.9637172 -0.8183743 0.04309569
## 865      1      1 1.9637172 -0.8183743 0.04309569
## 866      1      1 1.9637172 -0.8183743 0.04309569
## 867      1      1 1.9637172 -0.8183743 0.04309569
## 868      1      1 1.9637172 -0.8183743 0.04309569
## 869      1      1 1.9637172 -0.8183743 0.04309569
## 870      1      1 1.9637172 -0.8183743 0.04309569
## 871      1      1 1.9637172 -0.8183743 0.04309569
## 872      1      1 1.9637172 -0.8183743 0.04309569
## 873      1      1 1.9637172 -0.8183743 0.04309569
## 874      1      1 1.9637172 -0.8183743 0.04309569
## attr("assign")
## [1] 0 1 2 3 4
## attr("contrasts")
## attr("contrasts")$EggOrder
## [1] "contr.treatment"
##
## attr("msgScaleX")
## character(0)
```

Third Top Model

```
##run model
bm3 <- lmer(Size ~ EggOrder + ANCHL + BEUTI + SST + (1 | Observer) + (1 | Plot), data = df)

## look at model output and estimates
summary(bm3)

## Linear mixed model fit by REML ['lmerMod']
## Formula: Size ~ EggOrder + ANCHL + BEUTI + SST + (1 | Observer) + (1 |
##      Plot)
##      Data: df
##
## REML criterion at convergence: 9212.7
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -6.2323 -0.5763  0.0686  0.6646  2.3094
##
## Random effects:
##      Groups   Name      Variance Std.Dev.
##      Observer (Intercept)  110.5    10.51
##      Plot      (Intercept)   393.7    19.84
##      Residual              12384.7   111.29
## Number of obs: 753, groups:  Observer, 27; Plot, 8
##
## Fixed effects:
##              Estimate Std. Error t value
```

```
## (Intercept) 1883.766      10.440 180.433
## EggOrderEgg2 29.471      10.117  2.913
## ANCHL        23.916       7.467  3.203
## BEUTI        12.624       5.642  2.237
## SST         -2.913       6.967 -0.418
##
## Correlation of Fixed Effects:
##              (Intr) EggOE2 ANCHL  BEUTI
## EggOrdrEgg2 -0.213
## ANCHL        0.030 -0.016
## BEUTI        0.083 -0.021  0.438
## SST         0.112 -0.009 -0.518  0.162
```

```
coef(bm3)
```

```
## $Observer
##      (Intercept) EggOrderEgg2      ANCHL      BEUTI      SST
## AAY      1891.907      29.47095 23.91574 12.62421 -2.913246
## AJB      1881.994      29.47095 23.91574 12.62421 -2.913246
## AJD      1886.099      29.47095 23.91574 12.62421 -2.913246
## AML      1884.089      29.47095 23.91574 12.62421 -2.913246
## CAC      1878.427      29.47095 23.91574 12.62421 -2.913246
## CEH      1881.634      29.47095 23.91574 12.62421 -2.913246
## CEK      1883.272      29.47095 23.91574 12.62421 -2.913246
## CLE      1880.340      29.47095 23.91574 12.62421 -2.913246
## DLW      1880.529      29.47095 23.91574 12.62421 -2.913246
## EWW      1885.244      29.47095 23.91574 12.62421 -2.913246
## GRK      1883.196      29.47095 23.91574 12.62421 -2.913246
## JAH      1887.207      29.47095 23.91574 12.62421 -2.913246
## KMR      1885.164      29.47095 23.91574 12.62421 -2.913246
## KWB      1887.102      29.47095 23.91574 12.62421 -2.913246
## LAH      1883.771      29.47095 23.91574 12.62421 -2.913246
## MEJ      1881.681      29.47095 23.91574 12.62421 -2.913246
## MGB      1884.529      29.47095 23.91574 12.62421 -2.913246
## NAG      1879.407      29.47095 23.91574 12.62421 -2.913246
## PTL      1880.153      29.47095 23.91574 12.62421 -2.913246
## RER      1885.036      29.47095 23.91574 12.62421 -2.913246
## REW      1883.064      29.47095 23.91574 12.62421 -2.913246
## SAA      1892.769      29.47095 23.91574 12.62421 -2.913246
## SFC      1883.628      29.47095 23.91574 12.62421 -2.913246
## SJK      1886.354      29.47095 23.91574 12.62421 -2.913246
## SKT      1874.150      29.47095 23.91574 12.62421 -2.913246
## SLA      1884.664      29.47095 23.91574 12.62421 -2.913246
## SMC      1886.276      29.47095 23.91574 12.62421 -2.913246
##
## $Plot
##      (Intercept) EggOrderEgg2      ANCHL      BEUTI      SST
## APNC      1901.622      29.47095 23.91574 12.62421 -2.913246
## BH        1894.684      29.47095 23.91574 12.62421 -2.913246
## BT        1860.679      29.47095 23.91574 12.62421 -2.913246
## CC        1899.619      29.47095 23.91574 12.62421 -2.913246
## DO        1880.277      29.47095 23.91574 12.62421 -2.913246
## ESC       1882.730      29.47095 23.91574 12.62421 -2.913246
## LC        1873.241      29.47095 23.91574 12.62421 -2.913246
## WC        1877.278      29.47095 23.91574 12.62421 -2.913246
```

```
##
## attr("class")
## [1] "coef.mer"
```

```
ranef(bm3)
```

```
## $Observer
##      (Intercept)
## AAY      8.1408130
## AJB     -1.7723671
## AJD      2.3324821
## AML      0.3233991
## CAC     -5.3388210
## CEH     -2.1322987
## CEK     -0.4945298
## CLE     -3.4259606
## DLW     -3.2368660
## EWW      1.4775362
## GRK     -0.5699635
## JAH      3.4408482
## KMR      1.3975007
## KWB      3.3356245
## LAH      0.0051516
## MEJ     -2.0850697
## MGB      0.7629937
## NAG     -4.3594131
## PTL     -3.6128646
## RER      1.2704058
## REW     -0.7020500
## SAA      9.0024622
## SFC     -0.1381779
## SJK      2.5879189
## SKT     -9.6165109
## SLA      0.8976424
## SMC      2.5101145
```

```
##
## $Plot
##      (Intercept)
## APNC     17.855839
## BH       10.917564
## BT      -23.087343
## CC       15.852792
## DO       -3.488970
## ESC      -1.036457
## LC      -10.525551
## WC       -6.487875
```

```
##
## with conditional variances for "Observer" "Plot"
```

```
model.matrix(bm3)
```

```
##      (Intercept) EggOrderEgg2      ANCHL      BEUTI      SST
## 1              1              0 -0.4071777  0.6675426 -0.3804073
## 2              1              0 -0.4071777  0.6675426 -0.3804073
## 3              1              0 -0.4071777  0.6675426 -0.3804073
## 4              1              0 -0.8196345 -0.1856417 -0.2808295
```

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

## 329	1	0	-0.7585699	2.1517850	-0.3423867
## 330	1	0	-0.7585699	2.1517850	-0.3423867
## 331	1	0	-0.7585699	2.1517850	-0.3423867
## 332	1	0	-0.7585699	2.1517850	-0.3423867
## 333	1	0	-0.7585699	2.1517850	-0.3423867
## 334	1	0	-0.7585699	2.1517850	-0.3423867
## 335	1	0	-0.7585699	2.1517850	-0.3423867
## 336	1	0	-0.7585699	2.1517850	-0.3423867
## 337	1	0	-0.7585699	2.1517850	-0.3423867
## 338	1	0	-0.7585699	2.1517850	-0.3423867
## 339	1	0	-0.7585699	2.1517850	-0.3423867
## 340	1	0	-0.7585699	2.1517850	-0.3423867
## 341	1	0	-0.7585699	2.1517850	-0.3423867
## 342	1	0	-0.7585699	2.1517850	-0.3423867
## 343	1	0	-0.7585699	2.1517850	-0.3423867
## 344	1	0	-0.7585699	2.1517850	-0.3423867
## 345	1	0	-0.7585699	2.1517850	-0.3423867
## 346	1	0	-0.7585699	2.1517850	-0.3423867
## 347	1	0	-0.7585699	2.1517850	-0.3423867
## 348	1	0	-0.7585699	2.1517850	-0.3423867
## 349	1	0	-0.7585699	2.1517850	-0.3423867
## 350	1	0	-0.7585699	2.1517850	-0.3423867
## 351	1	0	-0.7585699	2.1517850	-0.3423867
## 352	1	0	-0.7585699	2.1517850	-0.3423867
## 353	1	0	-0.7585699	2.1517850	-0.3423867
## 354	1	0	-0.7585699	2.1517850	-0.3423867
## 355	1	0	-0.7585699	2.1517850	-0.3423867
## 356	1	0	0.9565296	-0.3861161	0.8416838
## 357	1	0	0.9565296	-0.3861161	0.8416838
## 358	1	0	0.9565296	-0.3861161	0.8416838
## 359	1	0	0.9565296	-0.3861161	0.8416838
## 360	1	0	0.9565296	-0.3861161	0.8416838
## 361	1	0	0.9565296	-0.3861161	0.8416838
## 362	1	0	0.9565296	-0.3861161	0.8416838
## 363	1	0	0.9565296	-0.3861161	0.8416838
## 364	1	0	0.9565296	-0.3861161	0.8416838
## 365	1	0	0.9565296	-0.3861161	0.8416838
## 366	1	0	0.9565296	-0.3861161	0.8416838
## 367	1	0	0.9565296	-0.3861161	0.8416838
## 368	1	0	0.9565296	-0.3861161	0.8416838
## 369	1	0	0.9565296	-0.3861161	0.8416838
## 370	1	0	0.9565296	-0.3861161	0.8416838
## 371	1	0	0.9565296	-0.3861161	0.8416838
## 372	1	0	0.9565296	-0.3861161	0.8416838
## 373	1	0	0.9565296	-0.3861161	0.8416838
## 374	1	0	0.9565296	-0.3861161	0.8416838
## 375	1	0	0.9565296	-0.3861161	0.8416838
## 376	1	0	0.9565296	-0.3861161	0.8416838
## 377	1	0	0.9565296	-0.3861161	0.8416838
## 378	1	0	0.9565296	-0.3861161	0.8416838
## 379	1	0	0.9565296	-0.3861161	0.8416838
## 380	1	0	0.9565296	-0.3861161	0.8416838
## 381	1	0	0.9565296	-0.3861161	0.8416838
## 382	1	0	0.9565296	-0.3861161	0.8416838

[illegible]

56

[illegible]

[illegible]

## 695	1	0	1.9637172	-0.8183743	0.2098175
## 696	1	0	1.9637172	-0.8183743	0.2098175
## 697	1	1	-0.4071777	0.6675426	-0.3804073
## 698	1	1	-0.4071777	0.6675426	-0.3804073
## 699	1	1	-0.4071777	0.6675426	-0.3804073
## 700	1	1	-0.4071777	0.6675426	-0.3804073
## 701	1	1	-0.8196345	-0.1856417	-0.2808295
## 702	1	1	-0.8196345	-0.1856417	-0.2808295
## 703	1	1	-0.8196345	-0.1856417	-0.2808295
## 704	1	1	-0.8196345	-0.1856417	-0.2808295
## 705	1	1	-0.8196345	-0.1856417	-0.2808295
## 706	1	1	-0.8196345	-0.1856417	-0.2808295
## 707	1	1	-0.8196345	-0.1856417	-0.2808295
## 708	1	1	-0.8196345	-0.1856417	-0.2808295
## 709	1	1	-0.8196345	-0.1856417	-0.2808295
## 710	1	1	-0.8196345	-0.1856417	-0.2808295
## 711	1	1	-0.8196345	-0.1856417	-0.2808295
## 712	1	1	-0.8196345	-0.1856417	-0.2808295
## 713	1	1	-0.8196345	-0.1856417	-0.2808295
## 714	1	1	-0.8196345	-0.1856417	-0.2808295
## 715	1	1	-0.8196345	-0.1856417	-0.2808295
## 716	1	1	-0.8196345	-0.1856417	-0.2808295
## 717	1	1	-0.8196345	-0.1856417	-0.2808295
## 718	1	1	-0.8196345	-0.1856417	-0.2808295
## 719	1	1	-0.8196345	-0.1856417	-0.2808295
## 720	1	1	-0.8196345	-0.1856417	-0.2808295
## 721	1	1	-0.8196345	-0.1856417	-0.2808295
## 722	1	1	-0.8196345	-0.1856417	-0.2808295
## 723	1	1	-0.8196345	-0.1856417	-0.2808295
## 724	1	1	-0.8196345	-0.1856417	-0.2808295
## 725	1	1	-0.8196345	-0.1856417	-0.2808295
## 726	1	1	-0.8196345	-0.1856417	-0.2808295
## 727	1	1	-0.7266228	-0.1902758	-0.7750975
## 728	1	1	-0.7266228	-0.1902758	-0.7750975
## 729	1	1	-0.7266228	-0.1902758	-0.7750975
## 730	1	1	-0.7266228	-0.1902758	-0.7750975
## 731	1	1	-0.7266228	-0.1902758	-0.7750975
## 732	1	1	-0.7266228	-0.1902758	-0.7750975
## 733	1	1	-0.7266228	-0.1902758	-0.7750975
## 734	1	1	-0.7266228	-0.1902758	-0.7750975
## 735	1	1	-0.7266228	-0.1902758	-0.7750975
## 736	1	1	-0.7266228	-0.1902758	-0.7750975
## 737	1	1	-0.7266228	-0.1902758	-0.7750975
## 738	1	1	-0.7266228	-0.1902758	-0.7750975
## 739	1	1	-0.7266228	-0.1902758	-0.7750975
## 740	1	1	-0.7266228	-0.1902758	-0.7750975
## 741	1	1	-0.7266228	-0.1902758	-0.7750975
## 742	1	1	-0.7848166	0.4290217	-1.7364758
## 743	1	1	-0.7848166	0.4290217	-1.7364758
## 744	1	1	-0.7848166	0.4290217	-1.7364758
## 745	1	1	-0.7848166	0.4290217	-1.7364758
## 746	1	1	-0.7848166	0.4290217	-1.7364758
## 747	1	1	-0.7848166	0.4290217	-1.7364758
## 748	1	1	-0.7848166	0.4290217	-1.7364758

## 749	1	1	-0.7848166	0.4290217	-1.7364758
## 750	1	1	-0.7848166	0.4290217	-1.7364758
## 751	1	1	-0.7848166	0.4290217	-1.7364758
## 752	1	1	-0.7848166	0.4290217	-1.7364758
## 753	1	1	-0.7848166	0.4290217	-1.7364758
## 754	1	1	-0.7848166	0.4290217	-1.7364758
## 755	1	1	-0.7848166	0.4290217	-1.7364758
## 756	1	1	-0.7848166	0.4290217	-1.7364758
## 757	1	1	-0.7848166	0.4290217	-1.7364758
## 758	1	1	-0.7848166	0.4290217	-1.7364758
## 759	1	1	-0.7848166	0.4290217	-1.7364758
## 760	1	1	-0.7848166	0.4290217	-1.7364758
## 761	1	1	-0.7848166	0.4290217	-1.7364758
## 762	1	1	-0.7848166	0.4290217	-1.7364758
## 763	1	1	-0.7848166	0.4290217	-1.7364758
## 764	1	1	-0.7848166	0.4290217	-1.7364758
## 765	1	1	-0.7848166	0.4290217	-1.7364758
## 766	1	1	-0.7848166	0.4290217	-1.7364758
## 767	1	1	-0.7848166	0.4290217	-1.7364758
## 768	1	1	-0.7585699	2.1517850	-0.3423867
## 769	1	1	-0.7585699	2.1517850	-0.3423867
## 770	1	1	-0.7585699	2.1517850	-0.3423867
## 771	1	1	-0.7585699	2.1517850	-0.3423867
## 772	1	1	-0.7585699	2.1517850	-0.3423867
## 773	1	1	-0.7585699	2.1517850	-0.3423867
## 774	1	1	-0.7585699	2.1517850	-0.3423867
## 775	1	1	-0.7585699	2.1517850	-0.3423867
## 776	1	1	-0.7585699	2.1517850	-0.3423867
## 777	1	1	-0.7585699	2.1517850	-0.3423867
## 778	1	1	-0.7585699	2.1517850	-0.3423867
## 779	1	1	-0.7585699	2.1517850	-0.3423867
## 780	1	1	-0.7585699	2.1517850	-0.3423867
## 781	1	1	-0.7585699	2.1517850	-0.3423867
## 782	1	1	-0.7585699	2.1517850	-0.3423867
## 783	1	1	-0.7585699	2.1517850	-0.3423867
## 784	1	1	-0.7585699	2.1517850	-0.3423867
## 785	1	1	-0.7585699	2.1517850	-0.3423867
## 786	1	1	-0.7585699	2.1517850	-0.3423867
## 787	1	1	-0.7585699	2.1517850	-0.3423867
## 788	1	1	-0.7585699	2.1517850	-0.3423867
## 789	1	1	0.9565296	-0.3861161	0.8416838
## 790	1	1	0.9565296	-0.3861161	0.8416838
## 791	1	1	0.9565296	-0.3861161	0.8416838
## 792	1	1	0.9565296	-0.3861161	0.8416838
## 793	1	1	0.9565296	-0.3861161	0.8416838
## 794	1	1	0.9565296	-0.3861161	0.8416838
## 795	1	1	0.9565296	-0.3861161	0.8416838
## 796	1	1	0.9565296	-0.3861161	0.8416838
## 797	1	1	0.9565296	-0.3861161	0.8416838
## 798	1	1	0.9565296	-0.3861161	0.8416838
## 799	1	1	0.9565296	-0.3861161	0.8416838
## 800	1	1	0.9565296	-0.3861161	0.8416838
## 801	1	1	0.9565296	-0.3861161	0.8416838
## 802	1	1	0.9565296	-0.3861161	0.8416838

```

## 803      1      1 0.9565296 -0.3861161 0.8416838
## 804      1      1 0.9565296 -0.3861161 0.8416838
## 805      1      1 0.9565296 -0.3861161 0.8416838
## 831      1      1 0.7911322 -1.3133684 0.8742729
## 832      1      1 0.7911322 -1.3133684 0.8742729
## 833      1      1 0.7911322 -1.3133684 0.8742729
## 834      1      1 0.7911322 -1.3133684 0.8742729
## 835      1      1 0.7911322 -1.3133684 0.8742729
## 836      1      1 0.7911322 -1.3133684 0.8742729
## 837      1      1 0.7911322 -1.3133684 0.8742729
## 838      1      1 0.7911322 -1.3133684 0.8742729
## 839      1      1 0.7911322 -1.3133684 0.8742729
## 840      1      1 0.7911322 -1.3133684 0.8742729
## 841      1      1 0.7911322 -1.3133684 0.8742729
## 842      1      1 0.7911322 -1.3133684 0.8742729
## 843      1      1 0.7911322 -1.3133684 0.8742729
## 844      1      1 0.7911322 -1.3133684 0.8742729
## 845      1      1 0.7911322 -1.3133684 0.8742729
## 846      1      1 0.7911322 -1.3133684 0.8742729
## 847      1      1 0.7911322 -1.3133684 0.8742729
## 848      1      1 0.7911322 -1.3133684 0.8742729
## 849      1      1 0.7911322 -1.3133684 0.8742729
## 850      1      1 0.7911322 -1.3133684 0.8742729
## 851      1      1 0.7911322 -1.3133684 0.8742729
## 852      1      1 0.7911322 -1.3133684 0.8742729
## 853      1      1 0.7911322 -1.3133684 0.8742729
## 854      1      1 0.7911322 -1.3133684 0.8742729
## 855      1      1 0.7911322 -1.3133684 0.8742729
## 856      1      1 0.7911322 -1.3133684 0.8742729
## 857      1      1 0.7911322 -1.3133684 0.8742729
## 858      1      1 0.7911322 -1.3133684 0.8742729
## 859      1      1 0.7911322 -1.3133684 0.8742729
## 860      1      1 0.7911322 -1.3133684 0.8742729
## 861      1      1 1.9637172 -0.8183743 0.2098175
## 862      1      1 1.9637172 -0.8183743 0.2098175
## 863      1      1 1.9637172 -0.8183743 0.2098175
## 864      1      1 1.9637172 -0.8183743 0.2098175
## 865      1      1 1.9637172 -0.8183743 0.2098175
## 866      1      1 1.9637172 -0.8183743 0.2098175
## 867      1      1 1.9637172 -0.8183743 0.2098175
## 868      1      1 1.9637172 -0.8183743 0.2098175
## 869      1      1 1.9637172 -0.8183743 0.2098175
## 870      1      1 1.9637172 -0.8183743 0.2098175
## 871      1      1 1.9637172 -0.8183743 0.2098175
## 872      1      1 1.9637172 -0.8183743 0.2098175
## 873      1      1 1.9637172 -0.8183743 0.2098175
## 874      1      1 1.9637172 -0.8183743 0.2098175
## attr("assign")
## [1] 0 1 2 3 4
## attr("contrasts")
## attr("contrasts")$EggOrder
## [1] "contr.treatment"
##
## attr("msgScaleX")

```

```
## character(0)
```

Model Diagnostics

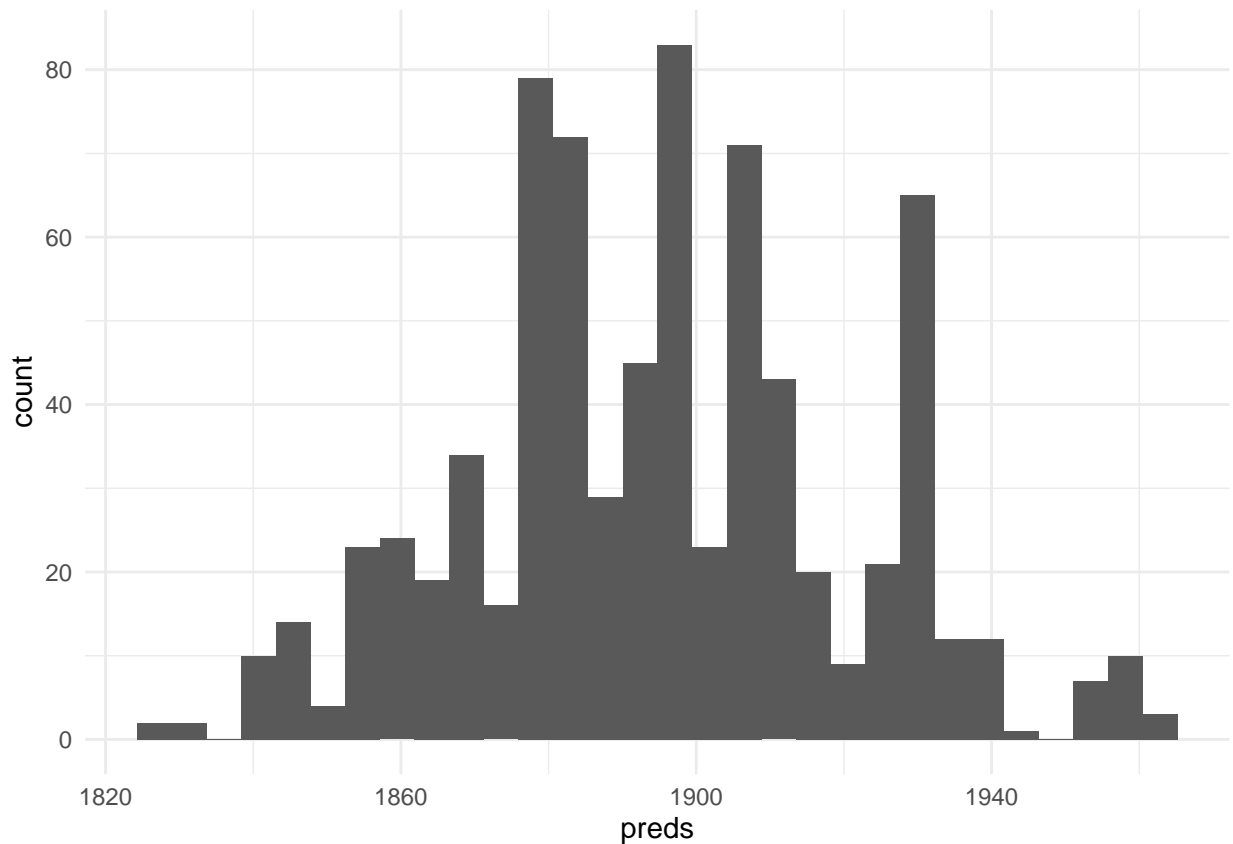
These diagnostics are done for the first top model only but can be repeated for the other 2 models.

Residuals/Fitted Plots

Histogram of Predicted Values

```
## extract predicted values and plot
preds <- predict(bm1)
ggplot() +
  geom_histogram(mapping = aes(preds)) +
  theme_minimal()
```

```
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```



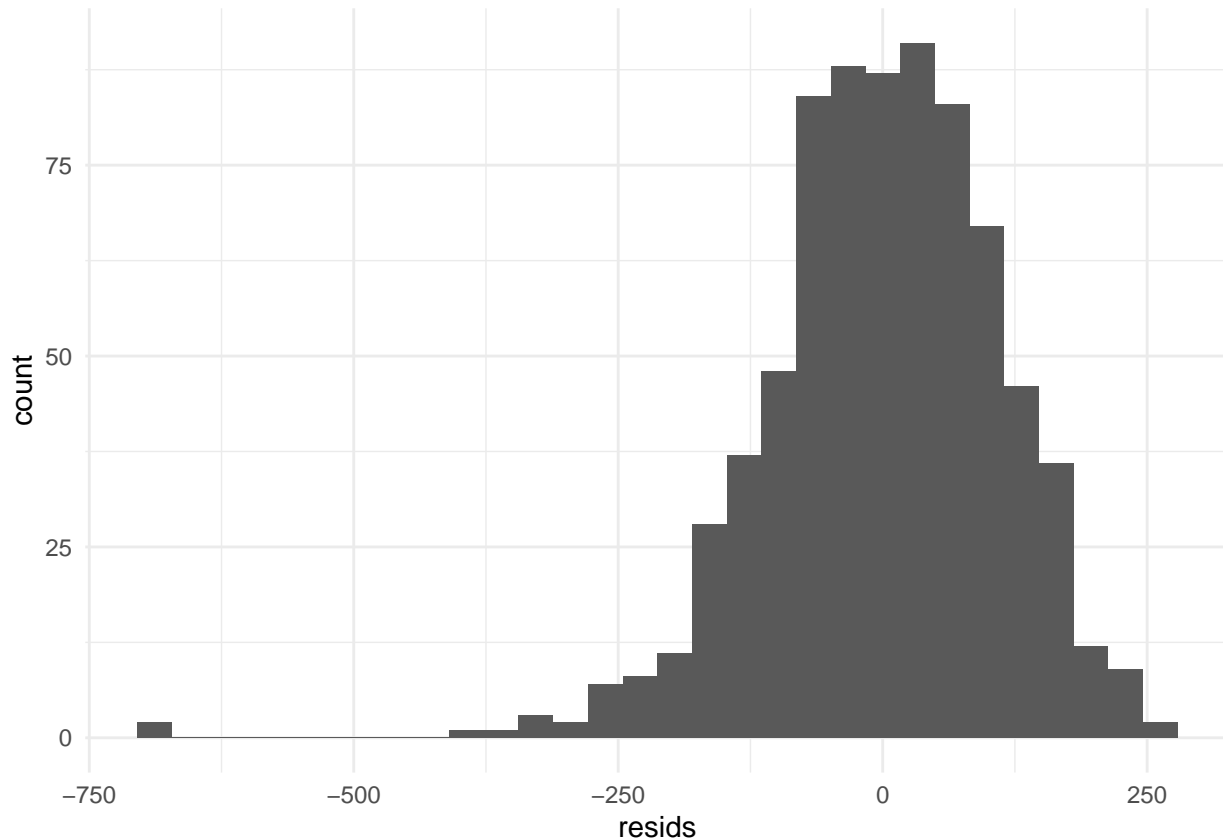
Since we assumed our data are normal, we want to see an approximately normal distribution of predicted values.

Histogram of Residuals

```
## extract residuals and plot
resids <- residuals(bm1)
ggplot() +
```

```
geom_histogram(mapping = aes(resids)) +
theme_minimal()
```

```
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```



Since we assumed our data are normal, we want to see an approximately normal distribution of residuals (differences between observed and predicted values of data).

Model Coefficients

```
## extract coeffs and random effects
coef(bm1) # this include fixed and random effects
```

```
## $Observer
##      (Intercept) EggOrderEgg2  BEUTI      NPG0      ONI
## AAY      1898.226    29.90976  11.1043 -22.75851  2.971758
## AJB      1887.490    29.90976  11.1043 -22.75851  2.971758
## AJD      1893.335    29.90976  11.1043 -22.75851  2.971758
## AML      1889.849    29.90976  11.1043 -22.75851  2.971758
## CAC      1882.309    29.90976  11.1043 -22.75851  2.971758
## CEH      1886.834    29.90976  11.1043 -22.75851  2.971758
## CEK      1888.526    29.90976  11.1043 -22.75851  2.971758
## CLE      1885.842    29.90976  11.1043 -22.75851  2.971758
## DLW      1886.151    29.90976  11.1043 -22.75851  2.971758
## EWW      1890.804    29.90976  11.1043 -22.75851  2.971758
## GRK      1888.900    29.90976  11.1043 -22.75851  2.971758
## JAH      1894.719    29.90976  11.1043 -22.75851  2.971758
```

```
## KMR      1890.363      29.90976 11.1043 -22.75851 2.971758
## KWB      1893.889      29.90976 11.1043 -22.75851 2.971758
## LAH      1889.444      29.90976 11.1043 -22.75851 2.971758
## MEJ      1887.006      29.90976 11.1043 -22.75851 2.971758
## MGB      1890.467      29.90976 11.1043 -22.75851 2.971758
## NAG      1885.550      29.90976 11.1043 -22.75851 2.971758
## PTL      1885.434      29.90976 11.1043 -22.75851 2.971758
## RER      1891.692      29.90976 11.1043 -22.75851 2.971758
## REW      1887.998      29.90976 11.1043 -22.75851 2.971758
## SAA      1900.548      29.90976 11.1043 -22.75851 2.971758
## SFC      1889.011      29.90976 11.1043 -22.75851 2.971758
## SJK      1893.045      29.90976 11.1043 -22.75851 2.971758
## SKT      1879.078      29.90976 11.1043 -22.75851 2.971758
## SLA      1890.326      29.90976 11.1043 -22.75851 2.971758
## SMC      1890.980      29.90976 11.1043 -22.75851 2.971758
##
## $Plot
##      (Intercept) EggOrderEgg2      BEUTI      NPGO      ONI
## APNC      1906.937      29.90976 11.1043 -22.75851 2.971758
## BH         1899.167      29.90976 11.1043 -22.75851 2.971758
## BT         1870.282      29.90976 11.1043 -22.75851 2.971758
## CC         1904.589      29.90976 11.1043 -22.75851 2.971758
## DO         1884.902      29.90976 11.1043 -22.75851 2.971758
## ESC        1888.654      29.90976 11.1043 -22.75851 2.971758
## LC         1877.422      29.90976 11.1043 -22.75851 2.971758
## WC         1884.437      29.90976 11.1043 -22.75851 2.971758
##
## attr(,"class")
## [1] "coef.mer"
```

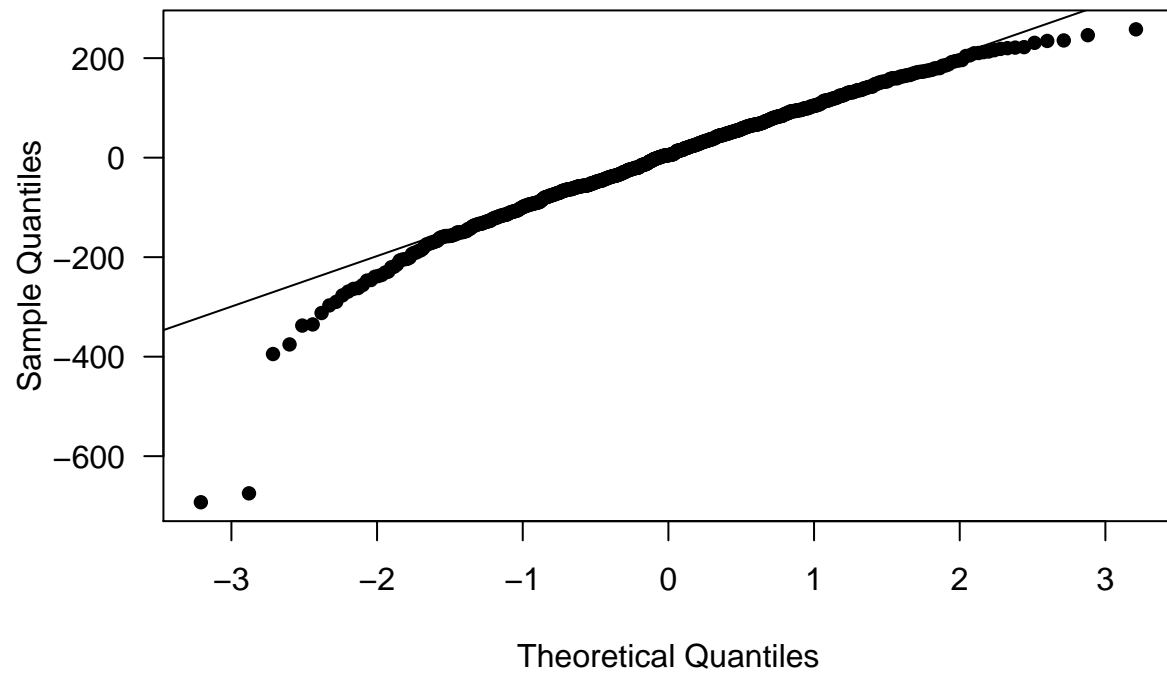
```
ranef_obs <- ranef(bm1)$Observer # observer random effect only
ranef_pl <- ranef(bm1)$Plot # plot random effect only
```

We can extract our model coefficients (for fixed and random effects) and look at them.

Q-Q Plots

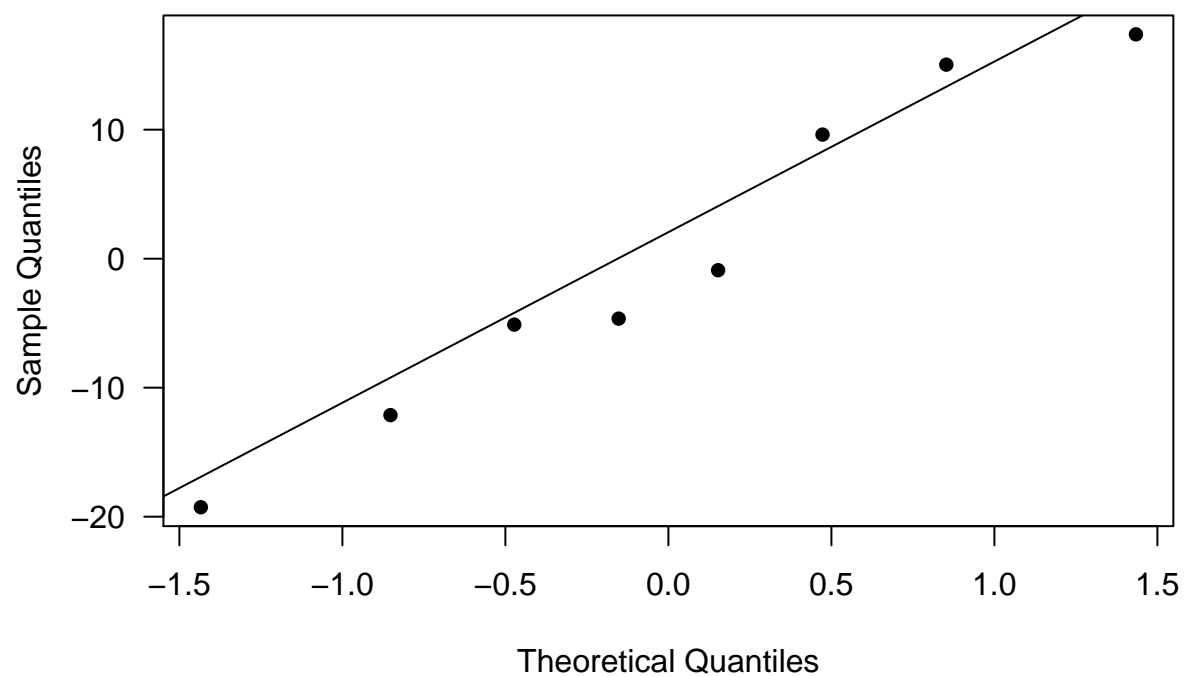
```
# qq resids
qqnorm(resids, main = "QQ plot (residuals)", las = 1, pch = 16)
qqline(resids)
```


QQ plot (residuals)

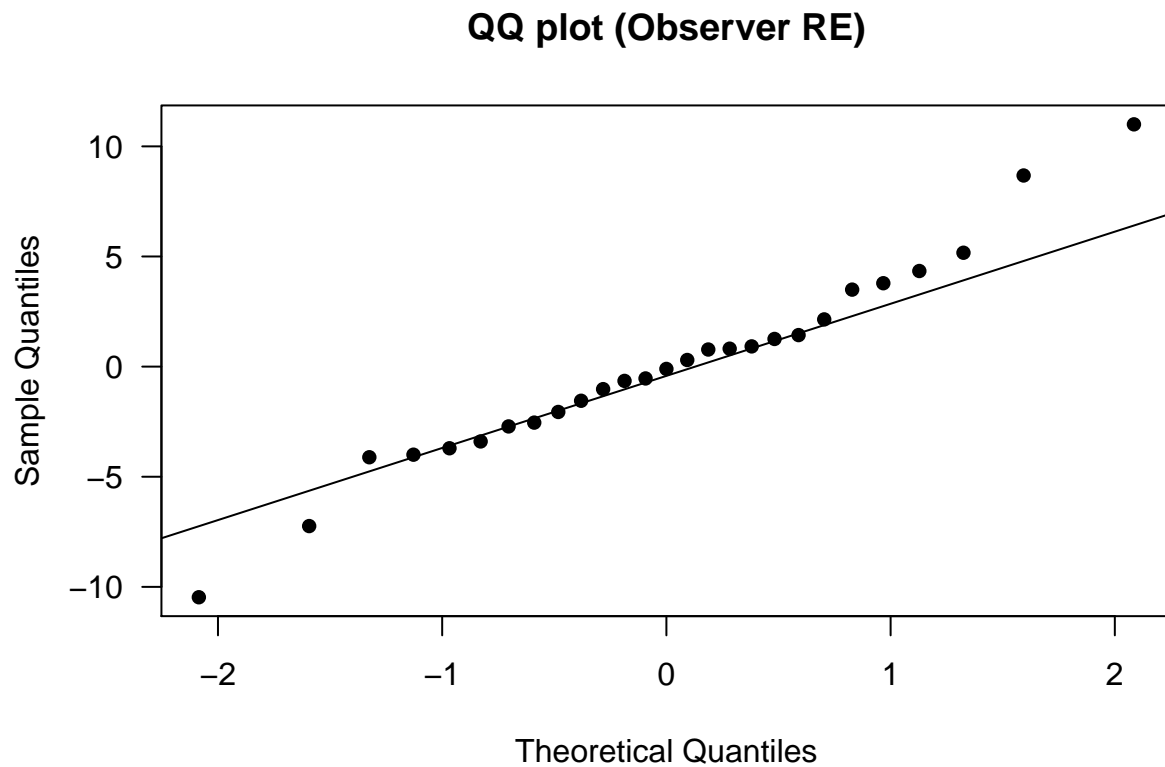


```
# qq Plot RE
qqnorm(unlist(ranef_pl), main = "QQ plot (Plot RE)", las = 1, pch = 16)
qqline(unlist(ranef_pl))
```

QQ plot (Plot RE)



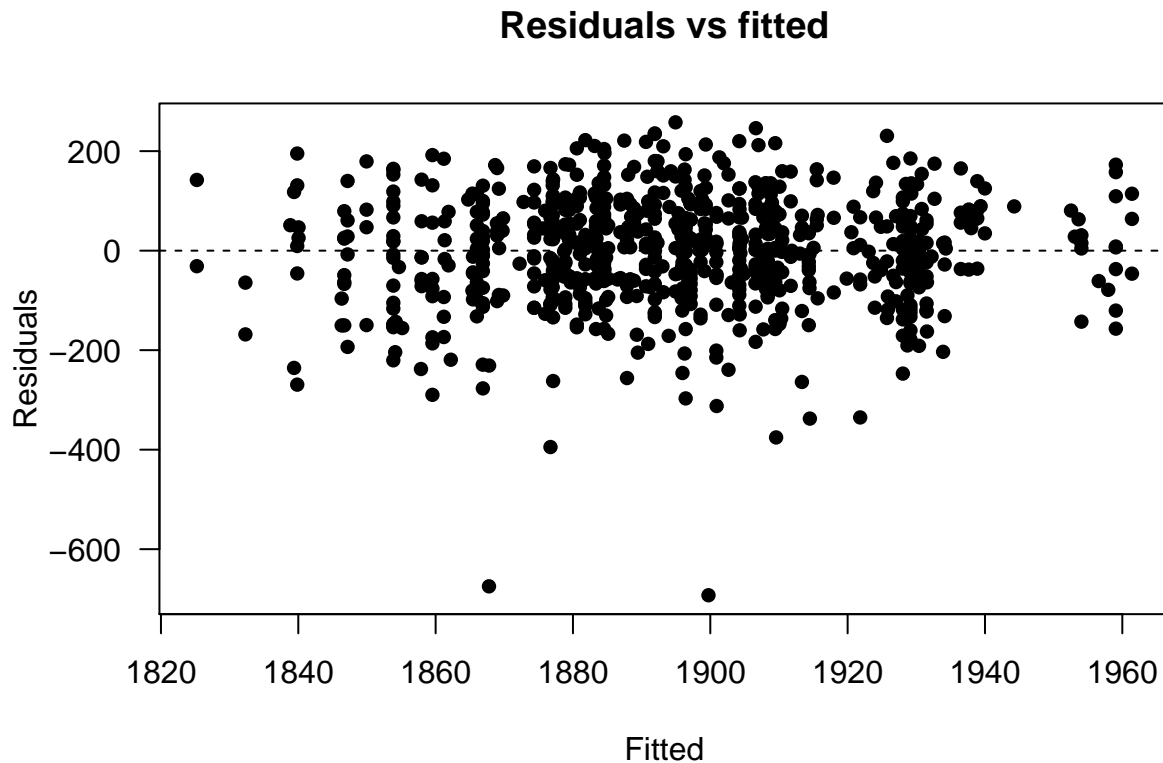
```
# qq Observer RE
qqnorm(unlist(ranef_obs), main = "QQ plot (Observer RE)", las = 1, pch = 16)
qqline(unlist(ranef_obs))
```



We want our points to fall approximately on the diagonal lines.

Fitted v Residuals

```
## plot residuals versus fitted values
yh <- fitted(bm1)
plot(yh, resid, las = 1, pch = 16,
     xlab = "Fitted", ylab = "Residuals",
     main = "Residuals vs fitted")
abline(h=0, lty = "dashed")
```



We assume our errors are normally distributed with constant variance. We want this plot to look like a scatterplot of points, without any evidence of trends.

Levene's test

We can formally test the assumption of homogenous variance via the Levene's Test, which compares the absolute values of the residuals among groups.

```
## split residuals into 2 groups
```

```
g1 <- resid[yh <= median(yh)]
```

```
g2 <- resid[yh > median(yh)]
```

```
## Levene's test
```

```
var.test(g1, g2)
```

```
##
```

```
## F test to compare two variances
```

```
##
```

```
## data: g1 and g2
```

```
## F = 1.0951, num df = 378, denom df = 373, p-value = 0.3791
```

```
## alternative hypothesis: true ratio of variances is not equal to 1
```

```
## 95 percent confidence interval:
```

```
## 0.8942338 1.3409987
```

```
## sample estimates:
```

```
## ratio of variances
```

```
## 1.09514
```

There is no justification to reject the null hypothesis that the residuals are equal. F is close to 1 and it is

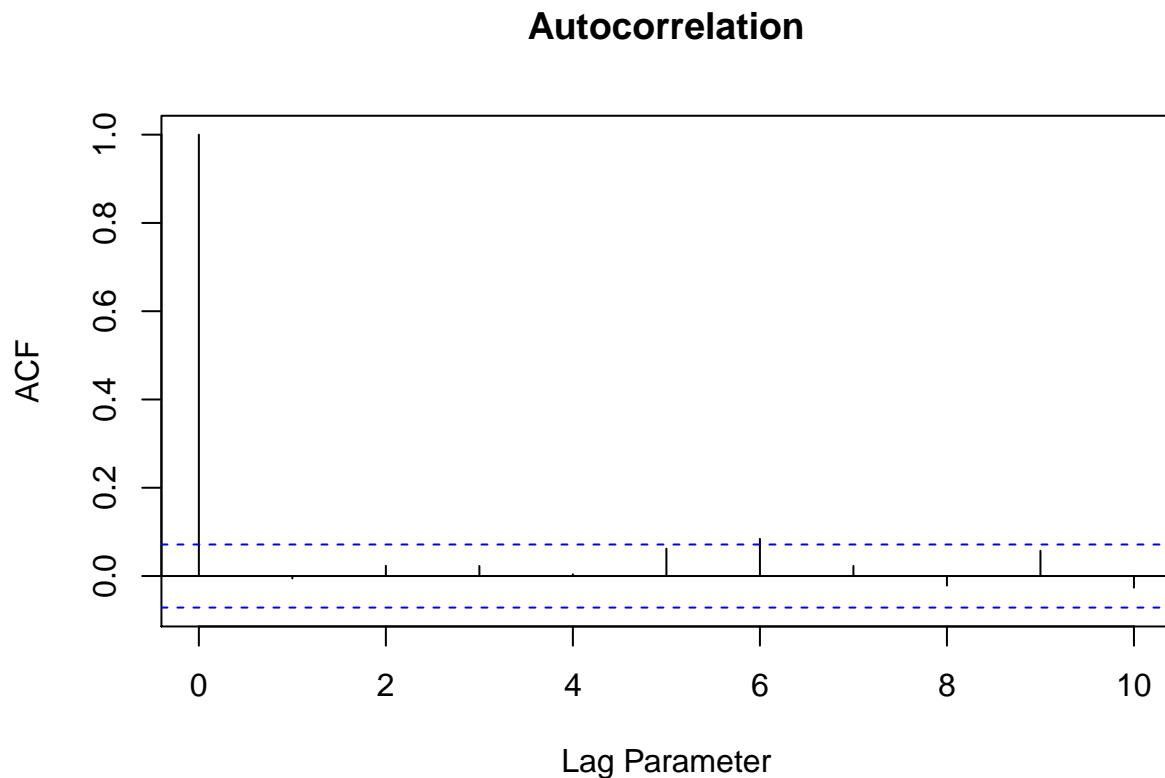
within the 95% confidence interval.

Autocorrelation

We also assume our errors are independent (e.g., not correlated). An ACF plot can be used to look for autocorrelation.

```
# calculate the ACF for lags between 1 and 10
autocorrelation <- acf(resids, lag.max= 10, plot = FALSE)

# Plot figure
plot(autocorrelation,
     main="Autocorrelation",
     xlab="Lag Parameter",
     ylab="ACF")
```



The first value with 0 lag will always be autocorrelated because it's stacked on itself. But after that, we want to see the values within the blue dotted lines. There does not appear to be autocorrelation in the residuals.

Goodness of Fit

Review with Sarah***

```
## Goodness of fit
## Pearson's X^2 statistic
true <- df$Size # 'true' egg size values
nn <- nrow(df) # number of observations
X2 <- sum((true - yh)^2/yh)
```

```
## Warning in true - yh: longer object length is not a multiple of shorter object
## length

## Warning in (true - yh)^2/yh: longer object length is not a multiple of shorter
## object length

degf <- nn-6 # 4 fixed effects, 2 random effects

## likelihood ratio test
pchisq(X2, df = degf, lower.tail = FALSE)

## [1] 0
```

Likelihood Ratio Tests

Review with Sarah***

```
## Likelihood ratio tests
## run model with plot RE only
bm_plot <- lmer(Size ~ EggOrder + BEUTI + NPGO + ONI + (1 | Plot), data = df)

## run model with obs RE only
bm_obs <- lmer(Size ~ EggOrder + BEUTI + NPGO + ONI + (1 | Observer), data = df)

## conduct an LRT to see if the variance of the Obs RE is contributing useful info
test_1 <- 2 * (logLik(bm1) - logLik(bm_obs))
pchisq(as.numeric(test_1), df = 1, lower.tail = FALSE)

## [1] 0.01460999
# There is support for inclusion of Obs as an RE

## check for contribution of Plot RE
test_2 <- 2 * (logLik(bm1) - logLik(bm_plot))
pchisq(as.numeric(test_2), df = 1, lower.tail = FALSE)

## [1] 1.504475e-316
# There is support for inclusion of Plot as an RE

## Bootstrapping to test for evidence against including multiple random effects in the same model
## set random seed
set.seed(514)

## fit null model with no RE's
nbm <- lm(Size ~ EggOrder + BEUTI + NPGO + ONI, data = df)

## calculate likelihood ratio (difference in log-likelihood)
lambda <- 2 * (logLik(bm1) - logLik(nbm))

## number of bootstrapped samples
nb <- 1000

## empty vector for storing LRT statistics
LRT_boot <- rep(NA, nb)

## bootstrapping
```


[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

```
## calculate approximate p-value
mean(LRT_boot > lambda)
```

```
## [1] 0.972
```

Identifying Outliers

https://qerm514.github.io/website/labs/week_03/diagnostics_and_errors.html#unusual_observations
https://qerm514.github.io/website/homework/week_03/hw_03_diagnostics_key.pdf Calculate the studentized residuals to look for outliers

```
## get studentized residuals
(stud_e <- rstudent(bm1))
```

##	1	2	3	4	5
##	1.1448036145	-1.5640157363	-0.5789843706	0.2366916785	-0.8749296749
##	6	7	8	9	10
##	0.4642884660	-0.5805845565	-1.1973921638	0.0362230507	-0.2851080407
##	11	12	13	14	15
##	1.7687502093	1.0340000625	0.4625459557	-0.8889883601	-0.7339670068
##	16	17	18	19	20
##	1.0470496740	-1.0393254067	-0.6501854445	-0.1026745676	1.5330707384
##	21	22	23	24	25
##	-1.0393254067	-0.4148915139	-0.9228444376	1.5103359358	0.2755537463
##	26	27	28	29	30

##	0.3551437585	-1.5403699714	-0.5885114457	0.8153896299	1.4665264884
##	31	32	33	34	35
##	0.1687745550	-1.3912408354	0.1687745550	-0.9481060288	-0.0708799832
##	36	37	38	39	40
##	0.5360277306	0.0602410298	-1.4282916484	0.2202149109	0.5563556709
##	41	42	43	44	45
##	-0.6310727947	0.9634740020	-0.7871348216	1.0584682792	-1.3654440878
##	46	47	48	49	50
##	0.7074548586	-0.0272718500	0.2176370529	0.2352590647	-0.5753017044
##	51	52	53	54	55
##	-0.6090370795	-0.1522387619	0.1893995080	0.7067871469	-0.2687997030
##	56	57	58	59	60
##	1.4867840796	-1.0530114116	0.6043127649	-1.9931806869	-1.3347185386
##	61	62	63	64	65
##	1.0738547884	1.3834161033	0.2637320135	-0.6890086565	-0.8287036453
##	66	67	68	69	70
##	-2.6235533886	-0.1305675286	-2.1618253418	1.0815143418	1.2975826306
##	71	72	73	74	75
##	1.1892347098	0.0847655391	-0.4490219895	-0.1428555311	0.8922599534
##	76	77	78	79	80
##	-1.3594074778	-2.4407582948	1.1884087017	-1.6871533435	-0.2989073874
##	81	82	83	84	85
##	0.5081846753	-1.3128295185	-0.6351985940	0.8765101744	-0.5150915641
##	86	87	88	89	90
##	-1.5765199822	1.7382257253	0.9719389746	0.4408703904	-0.2591251290
##	91	92	93	94	95
##	-0.5991540245	0.5508260812	1.2680995322	0.2573960331	0.7230543531
##	96	97	98	99	100
##	-0.1333082712	0.4473190294	0.2871893759	1.1059217567	0.5339345518
##	101	102	103	104	105
##	0.4270450036	0.2334373726	-0.9706817062	-1.7517847131	-0.0686373538
##	106	107	108	109	110
##	-1.8608465037	-0.4155356115	-1.3692683159	0.5000223669	-0.0611071312
##	111	112	113	114	115
##	-0.6636422612	0.5088555026	0.0247367098	0.5805587709	-0.3816289632
##	116	117	118	119	120
##	0.1903970310	1.5734068997	0.9658067535	-1.1217594471	-1.1521757684
##	121	122	123	124	125
##	1.4777973464	-0.3328805242	-1.8751082823	0.8675144978	-0.9737666966
##	126	127	128	129	130
##	1.0567060734	-0.7615045217	-0.2254772665	-0.8863950564	0.6183840689
##	131	132	133	134	135
##	-0.4380034085	-0.1263491241	1.6301934286	-0.5499485513	0.7747900693
##	136	137	138	139	140
##	1.5323038372	0.6670099221	-0.6056777582	-0.0173573062	0.5709578640
##	141	142	143	144	145
##	2.1233044583	1.0580769097	1.6265546069	-1.3569354518	0.4270689954
##	146	147	148	149	150
##	0.7470330915	0.2926009414	0.5016480856	0.1306506856	0.9005395278
##	151	152	153	154	155
##	0.9314557755	0.0409406429	-1.2315433994	1.0848037505	-0.4374624077
##	156	157	158	159	160
##	1.3771486023	-0.0755600232	-0.4686357286	-2.0844493710	0.4923659715
##	161	162	163	164	165

##	-0.4686357286	0.7405341206	-1.2146185700	-0.5287917965	0.2021896823
##	166	167	168	169	170
##	0.7581283851	-1.2076674497	0.8362686027	2.1326058780	-2.1027926539
##	171	172	173	174	175
##	2.0104672717	0.2397330282	-2.8333164210	-0.2379614533	1.3573320429
##	176	177	178	179	180
##	1.4483705865	0.7438564988	-0.8505616573	0.0236505703	-0.2014704023
##	181	182	183	184	185
##	1.9116751758	-2.1653065296	1.2920588901	0.5948141383	0.3106458384
##	186	187	188	189	190
##	-1.0408047327	0.4758855383	0.6976212523	0.3048590407	0.8391263767
##	191	192	193	194	195
##	-0.4989159632	0.9252738128	0.3343548305	0.2910515081	0.1258255679
##	196	197	198	199	200
##	-1.0177404038	-0.7121679789	0.4814719089	1.5595393948	0.8609587946
##	201	202	203	204	205
##	0.1931048960	-0.2476501352	-1.6576293776	-0.3474355629	1.7752536202
##	206	207	208	209	210
##	-0.1399709920	0.8376720881	0.9955671715	-0.9075840483	-0.3751176961
##	211	212	213	214	215
##	-0.1872894669	-2.3696057053	0.2077195390	1.2950794946	0.9034146782
##	216	217	218	219	220
##	0.0987439307	1.1320315214	-1.1597334578	0.1377973200	-1.4471193940
##	221	222	223	224	225
##	-0.7041648601	-0.9873629922	1.5052516617	-1.4269578299	-2.6887431129
##	226	227	228	229	230
##	0.5923331906	0.2768189686	-0.2634029431	-0.3128350134	1.1550982992
##	231	232	233	234	235
##	1.7548016583	0.9036741515	-0.0949038262	1.1817459790	0.6576057069
##	236	237	238	239	240
##	-2.5063383695	0.4172386410	0.3297819012	0.6007570515	0.2115634572
##	241	242	243	244	245
##	-3.5760077230	-1.7089617087	-0.8069082112	0.4738672160	1.1279843201
##	246	247	248	249	250
##	-0.7706511838	0.0898071669	-0.0260620930	-0.0260620930	1.5574235940
##	251	252	253	254	255
##	1.6738678957	-0.8443624214	-1.5731307008	-0.5498716917	1.2904972034
##	256	257	258	259	260
##	0.0615289747	-0.8859895737	-0.6460049661	-1.5100225614	-1.2043296464
##	261	262	263	264	265
##	0.4611620653	-0.4125199475	-0.3949706339	-0.1217485295	0.5344537654
##	266	267	268	269	270
##	-0.5160486958	0.0444866872	-0.5115927771	-0.8419316427	0.7954539735
##	271	272	273	274	275
##	0.0169948484	0.3575796491	-1.0226850696	-0.6754008337	0.1766948985
##	276	277	278	279	280
##	-0.9558717834	0.2608905031	0.8599920232	-0.2334191758	-0.3719345253
##	281	282	283	284	285
##	-0.1102918196	0.3265553768	0.0453128437	-0.1226528877	-0.5101150336
##	286	287	288	289	290
##	1.0430421693	-0.4266761085	-0.2594829236	-0.3329040968	0.1243426157
##	291	292	293	294	295
##	1.9864053407	-0.6509413571	0.3867687893	0.7060595404	0.0618255771
##	296	297	298	299	300

##	0.4713023625	-0.8668435726	0.4525165904	-0.9024462946	-1.1833314577
##	301	302	303	304	305
##	-0.5133259896	1.2854219956	0.5696431586	0.4628619950	1.4845470735
##	306	307	308	309	310
##	0.3351083997	-1.3623361557	0.9494361175	-0.1646495501	-0.8682909929
##	311	312	313	314	315
##	0.1737161770	0.0417074323	-1.1602526014	2.0152154829	0.2864906578
##	316	317	318	319	320
##	-0.5208947845	-0.9348850950	0.1423836827	1.4491702136	1.6306683429
##	321	322	323	324	325
##	1.4491702136	0.4061737285	0.8758638794	0.8617854239	0.4453356276
##	326	327	328	329	330
##	1.0937739100	0.8490286543	-0.5254695626	-0.5245626024	0.0610616131
##	331	332	333	334	335
##	0.4071155687	-0.9843316695	-0.2507985420	-1.0052897589	-0.5969242227
##	336	337	338	339	340
##	-0.3564969980	0.3175498833	-0.3386638953	0.2579431835	-0.2882644213
##	341	342	343	344	345
##	0.7802040825	-1.2410082496	1.9109134944	0.3066141985	-0.0221864670
##	346	347	348	349	350
##	0.2233755547	0.7289700921	-0.8217849455	-0.8283007760	-0.5411327156
##	351	352	353	354	355
##	-1.1540945845	-0.2914586780	-0.5823340400	-0.2065672828	-0.1263830890
##	356	357	358	359	360
##	-0.6672873043	-1.7309828127	-1.2472697578	0.0300329486	-0.4669361929
##	361	362	363	364	365
##	-0.2083372698	-1.9410042686	-0.2172143520	-0.0539443235	0.0113636879
##	366	367	368	369	370
##	-2.2308780368	-0.6888833232	0.6007307907	-0.7634530904	-0.8694325972
##	371	372	373	374	375
##	0.6509891366	0.6183827696	-0.8146202205	-0.1853433675	-0.8485863556
##	376	377	378	379	380
##	-0.1836269937	-0.3871527976	-0.7238782454	0.6834766267	0.1274256927
##	381	382	383	384	385
##	-0.4586846218	1.6030021783	0.1194161359	0.1758684997	0.0718421499
##	386	387	388	389	390
##	-1.5445395004	0.8953840857	-0.3247583087	0.4366567794	0.5314724628
##	391	392	393	394	395
##	-2.2311856495	0.2246027609	-0.8282117853	-0.3496597188	0.8662975773
##	396	397	398	399	400
##	0.8510709206	0.8671132910	0.8459953684	0.5642115853	-0.5091770743
##	401	402	403	404	405
##	1.3890101524	-2.1745481539	0.5873003479	0.9174108077	-1.1758618467
##	406	407	408	409	410
##	1.2399333328	0.6033129654	0.5661149174	-0.4172733252	-0.9717774052
##	411	412	413	414	415
##	-0.7181810639	0.1781070772	-0.3218631934	0.1330068600	-1.4709611656
##	416	417	418	419	420
##	-0.5759545090	-0.0464872125	-0.0023937435	-1.8449954370	-0.3021651114
##	421	422	423	424	425
##	0.6696324586	-1.8150479796	0.2096930261	0.6020307905	-0.0325044590
##	426	427	428	429	430
##	-0.9825076309	0.3055395522	-0.3284347450	0.1729009387	-0.1795548621
##	431	432	433	434	435

##	0.9654573908	-0.4958501186	0.1619416915	-0.6048874876	-0.0049999860
##	436	437	438	439	536
##	-1.0695580769	0.1230886038	0.0335394655	-0.1852536187	0.4055787275
##	537	538	539	540	541
##	-6.1271357675	0.0129474782	-0.5164817969	-1.3945874354	-0.4294338576
##	542	543	544	545	546
##	0.6629104363	-0.7233581477	-0.0131377273	-0.4734440309	-1.3456025776
##	547	548	549	550	551
##	0.8099122532	0.9510719449	0.4923708555	0.0041625011	1.1655813495
##	552	553	554	555	556
##	-0.3747095561	-0.4800507256	0.8548395360	-0.3139608458	-0.0010325460
##	557	558	559	560	561
##	0.8892977494	-1.3028675852	-0.8179567512	0.5895647168	-1.4333867489
##	562	563	564	565	566
##	-0.0931650890	-0.1895161597	0.5858436786	-1.4225048257	1.2609894744
##	567	568	569	570	571
##	-0.3413339938	-0.1558687452	0.2968916025	0.3610683782	-0.8494321562
##	572	573	574	575	576
##	0.2146368619	0.8654436008	-0.5321638710	0.0354152276	-0.5029905884
##	577	578	579	580	581
##	0.1343558324	-0.4946882539	0.6046557293	-0.1755706100	-1.3387013716
##	582	583	584	585	586
##	1.3826347021	0.4256759188	-0.5523376262	0.1488430362	1.8619548819
##	587	588	589	590	591
##	1.3809797167	2.3413123093	-0.7450425113	0.4942389945	0.0467030181
##	592	593	594	595	596
##	1.3538095137	1.2442378092	-0.5959999893	0.8468108875	0.0154766191
##	597	598	599	600	601
##	0.6564786503	-0.2253279748	-0.4413027666	-0.5316540301	0.7991956611
##	602	603	604	605	606
##	0.9258140483	0.8990686157	-0.2205901447	0.3276241430	-0.4385200125
##	607	608	609	610	611
##	0.4017437994	0.9972319655	0.4082518667	0.5238604527	1.1870505976
##	612	613	614	615	616
##	0.6376612422	-0.6650369135	0.0517720539	-0.1773138842	0.2642300581
##	617	618	619	620	621
##	1.9890167309	-0.6250933018	-0.7043470958	-0.9995832930	1.1256293332
##	622	623	624	625	626
##	-0.8008614082	0.3368869000	-0.4603502386	0.8474729160	0.8379790506
##	627	628	629	630	631
##	0.3860704600	1.9400177063	0.4796346330	0.1282916160	0.7625094067
##	632	633	634	635	636
##	0.2402230999	0.9385755893	-0.5518225303	1.5457919370	1.1935214955
##	637	638	639	640	641
##	-0.4275556275	0.4712540382	1.8469628151	1.3375951318	-0.5242314527
##	642	643	644	645	646
##	1.2134398836	-0.5103757210	1.2322615281	0.9436853383	1.5830899400
##	647	648	649	650	651
##	-0.2239857372	1.9517252170	-0.5716410334	-1.1069830147	1.6729384173
##	652	653	654	655	656
##	-0.5631032857	-0.1259565457	-0.3998958147	-1.6111716268	1.0727648455
##	657	658	659	660	661
##	-1.2615816337	-1.4230647840	0.6142254364	0.4896763970	-0.2120862760
##	662	663	664	665	666

##	0.3922971721	-1.2006691612	-0.6822484067	0.0024509945	0.7210374586
##	667	668	669	670	671
##	0.4064751496	0.1406956300	0.3780266794	-1.4473319188	0.4109933240
##	672	673	674	675	676
##	0.2149095281	-0.2113085434	0.4420632576	0.0004908585	0.6277603025
##	677	678	679	680	681
##	1.1827740871	1.5874601777	0.5554723041	-1.0877058094	0.7538899871
##	682	683	684	685	686
##	-0.2003100147	1.1300842007	-0.5099300229	-0.6927593208	0.3568644895
##	687	688	689	690	691
##	-0.5906507558	0.0498928826	-0.4689346725	-0.4942881646	-1.1212782553
##	692	693	694	695	696
##	-1.0562481618	1.0877588427	-0.1872883688	-0.9897729550	-0.9613631873
##	697	698	699	700	701
##	-3.4120958385	1.2293396682	-0.1801503222	1.0422070708	-2.0115471479
##	702	703	704	705	706
##	0.7371898265	-1.0950234020	-0.2925093043	-0.5174692223	0.0327991205
##	707	708	709	710	711
##	-0.4329359488	0.6737116923	1.1012377552	-1.4241671193	0.3648175016
##	712	713	714	715	716
##	0.3479965702	-1.8596079839	-0.3147686074	-0.0593664653	1.2318057301
##	717	718	719	720	721
##	0.6055076185	-0.3898136937	-0.3594600446	-0.6632214786	0.7541069218
##	722	723	724	725	726
##	-1.0540405513	-0.1305441460	0.7640613608	0.6300139514	0.0566939567
##	727	728	729	730	731
##	0.6085202762	1.2167892246	1.0423921303	1.0264173209	-0.6157312236
##	732	733	734	735	736
##	0.5302157673	1.7031786964	0.1032419248	-1.5418320845	1.9260340005
##	737	738	739	740	741
##	-0.5272212716	0.3923872597	1.2307542180	-3.0430826141	-0.0192267982
##	742	743	744	745	746
##	-3.0621913082	2.2357789861	0.9491989701	-0.2903401578	0.0455452665
##	747	748	749	750	751
##	0.2428787474	0.8061739351	0.7869302594	-1.1692884822	-0.2087548335
##	752	753	754	755	756
##	-0.3615318537	-0.8433204485	0.3880294206	-0.3735784825	1.3720301452
##	757	758	759	760	761
##	0.6826012946	0.9444189476	0.1994916520	-2.3209151935	-1.0411329889
##	762	763	764	765	766
##	-0.5718077206	-0.4709031003	0.4987717661	-0.4152659945	-0.6402511647
##	767	768	769	770	771
##	0.0509287887	0.3365572073	2.0992619543	0.6394267085	-1.1056301695
##	772	773	774	775	776
##	1.2704737734	0.5904844289	0.4436514312	-1.2312431658	0.9006359351
##	777	778	779	780	781
##	-0.6429258814	1.4446148682	0.7411987907	0.8113417151	0.4214768311
##	782	783	784	785	786
##	-2.4045822818	-0.0545025384	-0.1965434406	-0.3117783166	0.2873903366
##	787	788	789	790	791
##	-0.3292736271	-0.1133124290	1.3972296046	-0.5566771584	-0.7130131929
##	792	793	794	795	796
##	-1.2982606785	0.1368084677	0.2808884276	0.0398455654	-0.9254208162
##	797	798	799	800	801

```
## 0.6295058598 0.0390828546 0.2560076156 0.5410502608 0.7587435076
##      802      803      804      805      831
## -1.0772602484 -0.8376204897 0.7341167967 -0.5746014452 -1.7380516215
##      832      833      834      835      836
## -0.8190891837 -6.3041472261 -1.1924702416 -0.3352234409 1.1370713934
##      837      838      839      840      841
## 0.3057558645 -0.1088475083 0.1896800965 -0.5725335310 -0.2638435336
##      842      843      844      845      846
## 1.1765353152 -0.2522155539 0.1473858939 -0.0636867764 0.3351703104
##      847      848      849      850      851
## 1.4487195518 -1.0583772528 0.7493371254 0.6330729757 -0.3457785894
##      852      853      854      855      856
## 0.3154172503 1.3321780050 -0.3931883711 0.4363600946 -0.3572130976
##      857      858      859      860      861
## 1.4991958049 0.5112012304 0.6926103564 -0.6828803771 0.5790872336
##      862      863      864      865      866
## 1.0378652553 1.5642963686 0.9894503799 -0.4198780848 1.4284631954
##      867      868      869      870      871
## -1.0912421661 0.4116838519 0.5747837771 0.0751120138 -0.3362813899
##      872      873      874
## 0.8098663254 0.0592648103 -1.4200490005
```

```
## get sample size
n <- nrow(df)

## Bonferroni correction: alpha/n
alpha <- 0.05/n

## critical t value
degf <- n - length(coef(bm1))-1 # should be more due to REs?
t_crit <- qt(1 - alpha/2, degf)

## compare t_stud to t_crit
sum(stud_e > t_crit, na.rm = TRUE)
```

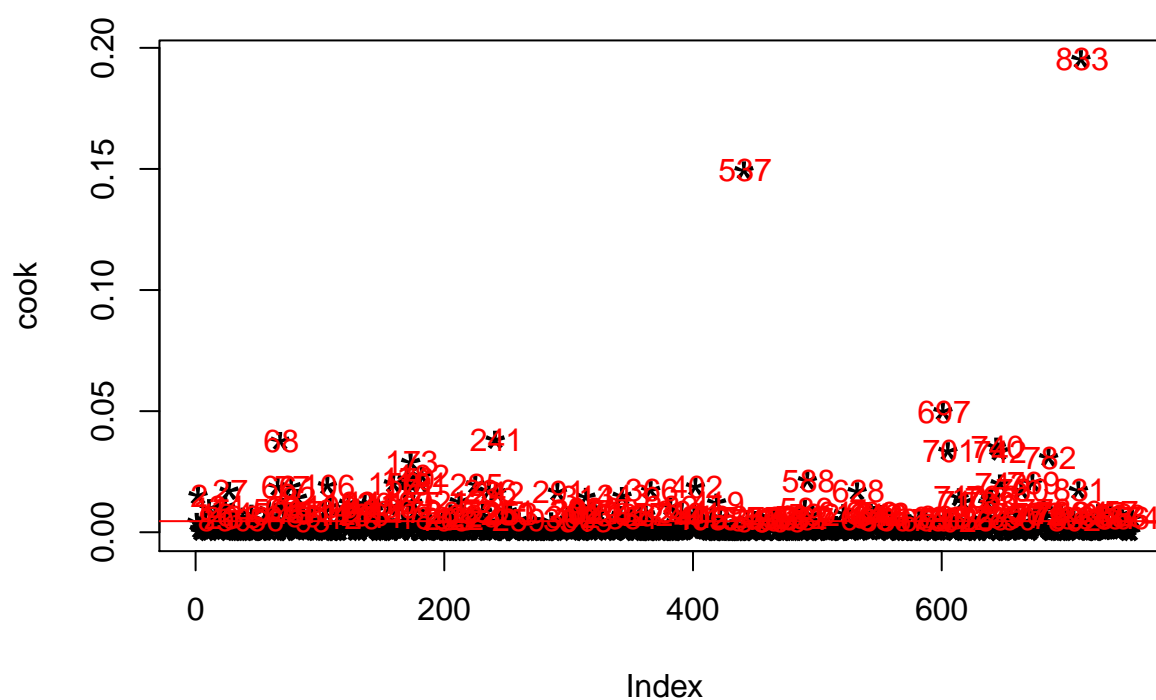
```
## [1] 0
```

Cook's Distance

```
## Cook's D
cook <- cooks.distance(bm1)

# Plot the Cook's Distance using the traditional 4/n criterion
sample_size <- nrow(df)
plot(cook, pch="*", cex=2, main="Influential Obs by Cooks distance") # plot cook's distance
abline(h = 4/sample_size, col="red") # add cutoff line
text(x=1:length(cook)+1, y=cook, labels=ifelse(cook>4/sample_size, names(cook),""), col="red") # add l
```

Influential Obs by Cooks distance



Confidence Intervals

```
# predict values
pred <- predict(bm1, re.form = NA)

# Bootstrap CI
boot <- bootMer(bm1, predict, nsim = 100, re.form = NA)
std.err <- apply(boot$t, 2, sd)
CI.lo <- pred - std.err*1.96
CI.hi <- pred + std.err*1.96

# Plot?
```

Prediction