

warmXtrophic Project: Herbivory Analyses

Kara Dobson

June 24, 2021

Load in and prepare data for analyses

```
# Clear all existing data
rm(list=ls())

#Load packages
library(tidyverse)
library(lmerTest)
library(olsrr)
library(predictmeans)
library(car)
library(fitdistrplus)
library(MASS)
library(pscl)
library(lmtest)
library(emmeans)

# Get data
#Sys.getenv("L1DIR")
L1_dir<-Sys.getenv("L1DIR")
#list.files(L1_dir)
herb <- read.csv(file.path(L1_dir, "herbivory/final_herbivory_L1.csv"))

# changing scale of years
herb$year1<-herb$year
herb$year[herb$year == 2015] <- 1
herb$year[herb$year == 2016] <- 2
herb$year[herb$year == 2017] <- 3
herb$year[herb$year == 2018] <- 4
herb$year[herb$year == 2019] <- 5
herb$year[herb$year == 2020] <- 6

# Remove NAs
herb <- herb[complete.cases(herb),]

# create dataframes for kbs and umbs only for plots with no insecticide
herb_kbs <- subset(herb, site == "kbs" & insecticide == "insects")
herb_umbs <- subset(herb, site == "umbs" & insecticide == "insects")

# only keep species that were recorded in both warmed and ambient plots
herb_kbs <- herb_kbs %>%
```

```

      group_by(species) %>%
      filter(all(c('warmed', 'ambient') %in% state))
herb_umbs <- herb_umbs %>%
      group_by(species) %>%
      filter(all(c('warmed', 'ambient') %in% state))

# checking to see if any species/state combos are all zeros
with(herb_kbs, table(species, state, p_eaten==0))

```

```

## , , = FALSE
##
##      state
## species ambient warmed
## Cest      78      39
## Eugr      33      65
## Hisp      27      11
## Hype       0       5
## Phpr      13      21
## Popr      19      14
## Soca     192     173
##
## , , = TRUE
##
##      state
## species ambient warmed
## Cest      64      42
## Eugr      44     103
## Hisp     165     117
## Hype       8      11
## Phpr      27      51
## Popr     183     176
## Soca     217     244

```

```

with(herb_umbs, table(species, state, p_eaten==0))

```

```

## , , = FALSE
##
##      state
## species ambient warmed
## Cape      10      14
## Cest     142     175
## Dasp      49      65
## Hype       9       8
## Poco       6      43
## Popr       1      11
## Posp      25      17
## Ptaq      27      39
## Ruac      80      98
##
## , , = TRUE
##
##      state
## species ambient warmed
## Cape      70      10

```

```
##      Cest      182    153
##      Dasp      131     87
##      Hype       55     40
##      Poco        6     21
##      Popr      107     85
##      Posp       23     47
##      Ptaq       29     65
##      Ruac       64    102
```

```
# number of observation per species/state combo (to find rare species)
herb_kbs %>% count(state, species)
```

```
## # A tibble: 14 x 3
## # Groups:   species [7]
##   species state      n
##   <chr>   <chr> <int>
## 1 Cest    ambient  142
## 2 Cest    warmed   81
## 3 Eugr    ambient   77
## 4 Eugr    warmed  168
## 5 Hisp    ambient  192
## 6 Hisp    warmed  128
## 7 Hype    ambient    8
## 8 Hype    warmed   16
## 9 Phpr    ambient   40
## 10 Phpr   warmed   72
## 11 Popr    ambient  202
## 12 Popr    warmed  190
## 13 Soca    ambient  409
## 14 Soca    warmed  417
```

```
herb_umbs %>% count(state, species)
```

```
## # A tibble: 18 x 3
## # Groups:   species [9]
##   species state      n
##   <chr>   <chr> <int>
## 1 Cape    ambient   80
## 2 Cape    warmed   24
## 3 Cest    ambient  324
## 4 Cest    warmed  328
## 5 Dasp    ambient  180
## 6 Dasp    warmed  152
## 7 Hype    ambient   64
## 8 Hype    warmed   48
## 9 Poco    ambient   12
## 10 Poco    warmed   64
## 11 Popr    ambient  108
## 12 Popr    warmed   96
## 13 Posp    ambient   48
## 14 Posp    warmed   64
## 15 Ptaq    ambient   56
## 16 Ptaq    warmed  104
## 17 Ruac    ambient  144
## 18 Ruac    warmed  200
```

```
# removing rare species from KBS
herb_kbs <- herb_kbs[!grepl("Hype",herb_kbs$species),]
herb_kbs %>% count(state, species)
```

```
## # A tibble: 12 x 3
## # Groups:   species [6]
##   species state     n
##   <chr>    <chr> <int>
## 1 Cest    ambient   142
## 2 Cest    warmed    81
## 3 Eogr    ambient    77
## 4 Eogr    warmed   168
## 5 Hisp    ambient   192
## 6 Hisp    warmed   128
## 7 Phpr    ambient    40
## 8 Phpr    warmed    72
## 9 Popr    ambient   202
## 10 Popr   warmed   190
## 11 Soca    ambient   409
## 12 Soca    warmed   417
```

```
# How much of the data is zeros?
```

```
100*sum(herb_kbs$p_eaten == 0)/nrow(herb_kbs) #68% - thats a lot! probably have to use a zero-inflated
```

```
## [1] 67.65817
```

```
# but I'll still check for normality & try some transformations below
```

```
100*sum(herb_umbs$p_eaten == 0)/nrow(herb_umbs) #61%
```

```
## [1] 60.92557
```

KBS

```
### determining distribution ###
```

```
# first, checking for normality
```

```
hist(herb_kbs$p_eaten)
```

```
#qqnorm(herb_kbs$p_eaten)
```

```
shapiro.test(herb_kbs$p_eaten)
```

```
#fit <- lm(p_eaten~state, data = herb_kbs)
```

```
#qqPlot(fit)
```

```
hist(herb_kbs$p_eaten[herb_kbs$state == "ambient"])
```

```
hist(herb_kbs$p_eaten[herb_kbs$state == "warmed"])
```

```
# not normal, attempting to transform data below
```

```
# log transform
```

```
herb_kbs$p_log <- log(herb_kbs$p_eaten+1)
```

```
hist(herb_kbs$p_log)
```

```
#qqnorm(herb_kbs$p_log)
```

```
shapiro.test(herb_kbs$p_log) # NAs - data contains 0s
```

```
# mean centering p_eaten
```

```
herb_kbs$p_scaled <- herb_kbs$p_log - mean(herb_kbs$p_log)
```

```
hist(herb_kbs$p_scaled)
```

```
hist(herb_kbs$p_scaled[herb_kbs$state == "ambient"])
```

```
hist(herb_kbs$p_scaled[herb_kbs$state == "warmed"])
```

```
#qqnorm(herb_kbs$p_scaled)
```

```
shapiro.test(herb_kbs$p_scaled)
# square root?
herb_kbs$p_sqrt <- sqrt(herb_kbs$p_eaten)
hist(herb_kbs$p_sqrt)
```

Transformations are a no-go

Going to try a zero-inflated model due to the excess number of zeros in the data

```
# mean and var of non-zero counts
herb_kbs %>%
  dplyr::filter(p_eaten != "0") %>%
  dplyr::summarize(mean_eaten = mean(p_eaten, na.rm=T), var_eaten = var(p_eaten, na.rm=T))

## `summarise()` ungrouping output (override with `.groups` argument)

## # A tibble: 6 x 3
##   species mean_eaten var_eaten
##   <chr>      <dbl>      <dbl>
## 1 Cest        9.41       156.
## 2 Eogr        6.60        66.3
## 3 Hisp       10.9       210.
## 4 Phpr       14.3       445.
## 5 Popr       17.8       455.
## 6 Soca        9.31       120.

# variance is also > mean, so can't be poisson
# I'll try zero-inflated negative binomial due to an excess of zeros

# zero-inflated negative binomial
# state as a fixed effect
k.m1 <- zeroinfl(p_eaten ~ state,
                dist = 'negbin',
                data = herb_kbs)
summary(k.m1)

##
## Call:
## zeroinfl(formula = p_eaten ~ state, data = herb_kbs, dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.3791 -0.3791 -0.3650 -0.1706 13.5408
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   1.8793     0.1236  15.208 < 2e-16 ***
## statearmed   -0.2704     0.1225  -2.208  0.0273 *
## Log(theta)   -1.1840     0.1778  -6.657 2.79e-11 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -0.2221     0.2274  -0.977  0.329
## statearmed    0.1209     0.1466   0.825  0.410
## ---
```

```

## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.3061
## Number of iterations in BFGS optimization: 14
## Log-likelihood: -3478 on 5 Df

# state and year as fixed effects
k.m2 <- zeroinfl(p_eaten ~ state + as.factor(year),
                dist = 'negbin',
                data = herb_kbs)
summary(k.m2)

##
## Call:
## zeroinfl(formula = p_eaten ~ state + as.factor(year), data = herb_kbs,
##          dist = "negbin")
##
## Pearson residuals:
##      Min      1Q   Median      3Q      Max
## -0.71839 -0.44651 -0.41647 -0.03154 24.51222
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.44518    0.11989   3.713 0.000205 ***
## statearmed     -0.16251    0.09187  -1.769 0.076902 .
## as.factor(year)2 1.42300    0.14034 10.140 < 2e-16 ***
## as.factor(year)3 2.21892    0.17820 12.452 < 2e-16 ***
## as.factor(year)4 2.19989    0.16157 13.616 < 2e-16 ***
## as.factor(year)5 2.18813    0.14669 14.917 < 2e-16 ***
## as.factor(year)6 -0.51223    0.23018  -2.225 0.026058 *
## Log(theta)     -0.25988    0.09587  -2.711 0.006716 **
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -9.6472    70.3916  -0.137  0.891
## statearmed      0.1375     0.1138   1.208  0.227
## as.factor(year)2 9.9493    70.3907   0.141  0.888
## as.factor(year)3 10.0800    70.3911   0.143  0.886
## as.factor(year)4 10.5992    70.3912   0.151  0.880
## as.factor(year)5 10.0197    70.3910   0.142  0.887
## as.factor(year)6 9.4078     70.3904   0.134  0.894
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.7711
## Number of iterations in BFGS optimization: 32
## Log-likelihood: -3324 on 15 Df

# state and growth habit as fixed effects
k.m3 <- zeroinfl(p_eaten ~ state + growth_habit,
                dist = 'negbin',
                data = herb_kbs)
summary(k.m3)

##
## Call:

```

```
## zeroinfl(formula = p_eaten ~ state + growth_habit, data = herb_kbs, dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.4727 -0.4510 -0.2344 -0.1774 12.4666
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      2.0326    0.2656   7.653 1.96e-14 ***
## statearmed       -0.2884    0.1177  -2.451  0.0143 *
## growth_habitForb -0.1991    0.2617  -0.761  0.4467
## growth_habitGraminoid 0.5203    0.3186   1.633  0.1025
## Log(theta)      -1.0808    0.1624  -6.654 2.85e-11 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      1.40009    0.22518   6.218 5.05e-10 ***
## statearmed        0.23661    0.16988   1.393  0.164
## growth_habitForb  -2.49970    0.30642  -8.158 3.42e-16 ***
## growth_habitGraminoid -0.07318    0.24395  -0.300  0.764
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.3393
## Number of iterations in BFGS optimization: 16
## Log-likelihood: -3340 on 9 Df
```

```
# state, growth habit, and year as fixed effects
k.m4 <- zeroinfl(p_eaten ~ state + growth_habit + as.factor(year),
  dist = 'negbin',
  data = herb_kbs)
summary(k.m4)
```

```
##
## Call:
## zeroinfl(formula = p_eaten ~ state + growth_habit + as.factor(year),
##      data = herb_kbs, dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.7411 -0.4548 -0.2839 -0.1254 25.0060
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      0.16585    0.24482   0.677  0.4981
## statearmed       -0.21211    0.08821  -2.405  0.0162 *
## growth_habitForb  0.32882    0.19555   1.682  0.0927 .
## growth_habitGraminoid 1.32892    0.24771   5.365 8.10e-08 ***
## as.factor(year)2    1.04559    0.16519   6.330 2.46e-10 ***
## as.factor(year)3    2.03927    0.19241  10.598 < 2e-16 ***
## as.factor(year)4    2.21073    0.17883  12.362 < 2e-16 ***
## as.factor(year)5    2.21518    0.17006  13.026 < 2e-16 ***
## as.factor(year)6   -0.47038    0.23756  -1.980  0.0477 *
## Log(theta)      -0.13936    0.10227  -1.363  0.1730
##
```

```
## Zero-inflation model coefficients (binomial with logit link):
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)      -1.5792      4.0319  -0.392   0.695
## statearmed         0.1768      0.1297   1.363   0.173
## growth_habitForb  -2.2263      0.2111 -10.544 <2e-16 ***
## growth_habitGraminoid 0.1036      0.2330   0.445   0.656
## as.factor(year)2    2.8669      4.0071   0.715   0.474
## as.factor(year)3    4.1002      4.0268   1.018   0.309
## as.factor(year)4    3.7977      4.0245   0.944   0.345
## as.factor(year)5    3.2469      4.0215   0.807   0.419
## as.factor(year)6    3.6623      4.0155   0.912   0.362
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.8699
## Number of iterations in BFGS optimization: 30
## Log-likelihood: -3155 on 19 Df

# interaction between state and growth habit as fixed effects
k.m5 <- zeroinfl(p_eaten ~ state * growth_habit,
                dist = 'negbin',
                data = herb_kbs)
summary(k.m5)

##
## Call:
## zeroinfl(formula = p_eaten ~ state * growth_habit, data = herb_kbs, dist = "negbin")
##
## Pearson residuals:
##      Min       1Q   Median       3Q      Max
## -0.4762 -0.4463 -0.2255 -0.1741 12.0804
##
## Count model coefficients (negbin with log link):
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)      1.99378    0.31158   6.399 1.56e-10 ***
## statearmed       -0.17091    0.55875  -0.306   0.760
## growth_habitForb -0.17585    0.31430  -0.559   0.576
## growth_habitGraminoid 0.58567    0.41126   1.424   0.154
## statearmed:growth_habitForb -0.09785    0.57356  -0.171   0.865
## statearmed:growth_habitGraminoid -0.18276    0.67908  -0.269   0.788
## Log(theta)      -1.09259    0.16783  -6.510 7.51e-11 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)      1.2872     0.2580   4.989 6.06e-07 ***
## statearmed         0.5590     0.4219   1.325   0.185
## growth_habitForb  -2.4733     0.3930  -6.293 3.11e-10 ***
## growth_habitGraminoid 0.1884     0.3153   0.597   0.550
## statearmed:growth_habitForb -0.2041     0.4952  -0.412   0.680
## statearmed:growth_habitGraminoid -0.6347     0.5122  -1.239   0.215
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.3353
## Number of iterations in BFGS optimization: 19
```



```
## Log-likelihood: -3339 on 13 Df
# interaction between state and growth habit as fixed effects, plus year
k.m6 <- zeroinfl(p_eaten ~ state * growth_habit + as.factor(year),
  dist = 'negbin',
  data = herb_kbs)
summary(k.m6)

##
## Call:
## zeroinfl(formula = p_eaten ~ state * growth_habit + as.factor(year),
##   data = herb_kbs, dist = "negbin")
##
## Pearson residuals:
##      Min       1Q   Median       3Q      Max
## -0.7380 -0.4559 -0.2932 -0.1375  24.8208
##
## Count model coefficients (negbin with log link):
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      0.158396   0.267077    0.593   0.5531
## statearmed       -0.157970   0.397350   -0.398   0.6910
## growth_habitForb  0.312405   0.228490    1.367   0.1715
## growth_habitGraminoid 1.546411   0.305612    5.060 4.19e-07 ***
## as.factor(year)2    1.041654   0.164111    6.347 2.19e-10 ***
## as.factor(year)3    2.069583   0.192773   10.736 < 2e-16 ***
## as.factor(year)4    2.216604   0.177985   12.454 < 2e-16 ***
## as.factor(year)5    2.207357   0.169311   13.037 < 2e-16 ***
## as.factor(year)6   -0.487322   0.237224   -2.054   0.0399 *
## statearmed:growth_habitForb 0.001971   0.407693    0.005   0.9961
## statearmed:growth_habitGraminoid -0.502881   0.481793   -1.044   0.2966
## Log(theta)       -0.129593   0.101681   -1.275   0.2025
##
## Zero-inflation model coefficients (binomial with logit link):
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      -1.5129     3.3120   -0.457   0.648
## statearmed        0.5501     0.4002    1.375   0.169
## growth_habitForb  -2.1203     0.2563  -8.272 <2e-16 ***
## growth_habitGraminoid 0.3579     0.2981    1.201   0.230
## as.factor(year)2    2.6836     3.2840    0.817   0.414
## as.factor(year)3    3.9207     3.3032    1.187   0.235
## as.factor(year)4    3.6121     3.3007    1.094   0.274
## as.factor(year)5    3.0515     3.2981    0.925   0.355
## as.factor(year)6    3.4608     3.2951    1.050   0.294
## statearmed:growth_habitForb -0.3307     0.4308   -0.767   0.443
## statearmed:growth_habitGraminoid -0.6483     0.4826   -1.343   0.179
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.8785
## Number of iterations in BFGS optimization: 33
## Log-likelihood: -3153 on 23 Df
# interaction between state, growth habit, and year (year as a factor wouldn't work - non-finite value)
k.m7 <- zeroinfl(p_eaten ~ state * growth_habit * year,
  dist = 'negbin',
```

```

data = herb_kbs)
summary(k.m7)

##
## Call:
## zeroinfl(formula = p_eaten ~ state * growth_habit * year, data = herb_kbs,
##   dist = "negbin")
##
## Pearson residuals:
##      Min      1Q   Median      3Q      Max
## -0.56843 -0.40005 -0.26576 -0.08172 11.76600
##
## Count model coefficients (negbin with log link):
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -0.8797     1.0092  -0.872  0.38339
## statearmed       0.3356     1.9062   0.176  0.86027
## growth_habitForb 1.6557     1.0293   1.609  0.10771
## growth_habitGraminoid 3.8580     1.1805   3.268  0.00108 **
## year            0.6947     0.2419   2.872  0.00408 **
## statearmed:growth_habitForb -0.7913     1.9309  -0.410  0.68197
## statearmed:growth_habitGraminoid -3.5563     2.3229  -1.531  0.12577
## statearmed:year    -0.1490     0.4341  -0.343  0.73144
## growth_habitForb:year -0.3554     0.2502  -1.421  0.15546
## growth_habitGraminoid:year -0.8110     0.3266  -2.483  0.01303 *
## statearmed:growth_habitForb:year 0.1748     0.4428   0.395  0.69291
## statearmed:growth_habitGraminoid:year 1.3010     0.6935   1.876  0.06065 .
## Log(theta)      -0.8096     0.1439  -5.625 1.86e-08 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)     2.1748     0.9173   2.371  0.01774 *
## statearmed       1.6455     1.8303   0.899  0.36863
## growth_habitForb -4.4783     1.0629  -4.213 2.52e-05 ***
## growth_habitGraminoid -2.4410     1.0628  -2.297  0.02164 *
## year            -0.2332     0.2205  -1.058  0.29018
## statearmed:growth_habitForb -2.3182     2.0202  -1.148  0.25117
## statearmed:growth_habitGraminoid -2.7263     2.0179  -1.351  0.17669
## statearmed:year    -0.2565     0.4253  -0.603  0.54650
## growth_habitForb:year 0.6463     0.2452   2.636  0.00840 **
## growth_habitGraminoid:year 0.8538     0.2888   2.956  0.00311 **
## statearmed:growth_habitForb:year 0.4303     0.4575   0.941  0.34693
## statearmed:growth_habitGraminoid:year 0.6680     0.5235   1.276  0.20193
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.445
## Number of iterations in BFGS optimization: 38
## Log-likelihood: -3266 on 25 Df

# state and origin as fixed effects
k.m8 <- zeroinfl(p_eaten ~ state + origin,
  dist = 'negbin',
  data = herb_kbs)
summary(k.m8)

```

```
##
## Call:
## zeroinfl(formula = p_eaten ~ state + origin, data = herb_kbs, dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.4643 -0.4335 -0.2957 -0.1633 12.6798
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   1.9766     0.2748   7.194 6.30e-13 ***
## statearmed    -0.2265     0.1206  -1.878  0.0604 .
## originExotic   0.1252     0.2845   0.440  0.6599
## originNative  -0.2241     0.2702  -0.830  0.4068
## Log(theta)    -1.1606     0.1757  -6.606 3.94e-11 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   1.2988     0.2396   5.420 5.95e-08 ***
## statearmed     0.3956     0.1704   2.322  0.0202 *
## originExotic  -1.0725     0.2320  -4.623 3.79e-06 ***
## originNative  -2.5301     0.3507  -7.215 5.39e-13 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.3133
## Number of iterations in BFGS optimization: 17
## Log-likelihood: -3399 on 9 Df
# state, origin, and year as fixed effects
k.m9 <- zeroinfl(p_eaten ~ state + origin + as.factor(year),
  dist = 'negbin',
  data = herb_kbs)
summary(k.m9)
```

```
##
## Call:
## zeroinfl(formula = p_eaten ~ state + origin + as.factor(year), data = herb_kbs,
##      dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.7219 -0.4221 -0.3205 -0.1137 24.9342
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   0.14561    0.23573   0.618  0.53677
## statearmed    -0.15828    0.08981  -1.762  0.07800 .
## originExotic   0.61304    0.21480   2.854  0.00432 **
## originNative   0.31058    0.19804   1.568  0.11682
## as.factor(year)2 1.30567    0.15286   8.542 < 2e-16 ***
## as.factor(year)3 2.06704    0.18635  11.092 < 2e-16 ***
## as.factor(year)4 2.12925    0.16958  12.556 < 2e-16 ***
## as.factor(year)5 2.22129    0.15866  14.000 < 2e-16 ***
## as.factor(year)6 -0.48355    0.23247  -2.080  0.03752 *
```

```

## Log(theta)      -0.15767    0.09422  -1.673  0.09426 .
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -1.0720     0.8557  -1.253 0.210267
## statewarmed      0.2822     0.1211   2.331 0.019764 *
## originExotic    -0.7190     0.2086  -3.446 0.000568 ***
## originNative    -2.1256     0.2095 -10.146 < 2e-16 ***
## as.factor(year)2  2.4214     0.8309   2.914 0.003565 **
## as.factor(year)3  3.2969     0.8462   3.896 9.77e-05 ***
## as.factor(year)4  3.1316     0.8380   3.737 0.000186 ***
## as.factor(year)5  2.7791     0.8364   3.323 0.000891 ***
## as.factor(year)6  2.9608     0.8787   3.370 0.000753 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.8541
## Number of iterations in BFGS optimization: 27
## Log-likelihood: -3229 on 19 Df

# interaction between state and origin as fixed effects
k.m10 <- zeroinfl(p_eaten ~ state * origin,
                  dist = 'negbin',
                  data = herb_kbs)
summary(k.m10)

##
## Call:
## zeroinfl(formula = p_eaten ~ state * origin, data = herb_kbs, dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.4616 -0.4373 -0.2958 -0.1653 12.6601
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    1.96897    0.31730   6.205 5.46e-10 ***
## statewarmed    -0.17160    0.56688  -0.303   0.762
## originExotic    0.13693    0.34136   0.401   0.688
## originNative   -0.19651    0.32341  -0.608   0.543
## statewarmed:originExotic -0.04501    0.61515  -0.073   0.942
## statewarmed:originNative -0.07525    0.58560  -0.129   0.898
## Log(theta)    -1.14400    0.17332  -6.601 4.09e-11 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    1.2554     0.2632   4.770 1.84e-06 ***
## statewarmed      0.5612     0.4252   1.320 0.186883
## originExotic    -1.0195     0.2859  -3.567 0.000362 ***
## originNative    -2.3913     0.4097  -5.837 5.31e-09 ***
## statewarmed:originExotic -0.1565     0.4806  -0.326 0.744684
## statewarmed:originNative -0.2694     0.5133  -0.525 0.599748
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##

```

```

## Theta = 0.3185
## Number of iterations in BFGS optimization: 18
## Log-likelihood: -3399 on 13 Df
# interaction between state and origin as fixed effects, plus year
k.m11 <- zeroinfl(p_eaten ~ state * origin + as.factor(year),
  dist = 'negbin',
  data = herb_kbs)
summary(k.m11)

##
## Call:
## zeroinfl(formula = p_eaten ~ state * origin + as.factor(year), data = herb_kbs,
##   dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.7179 -0.4281 -0.3252 -0.1216 24.4702
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.14984    0.26059   0.575  0.56528
## statewarmed   -0.18256    0.40246  -0.454  0.65012
## originExotic   0.65684    0.25133   2.613  0.00896 **
## originNative   0.28071    0.23332   1.203  0.22892
## as.factor(year)2  1.30898    0.15326   8.541 < 2e-16 ***
## as.factor(year)3  2.08185    0.18733  11.113 < 2e-16 ***
## as.factor(year)4  2.12297    0.16979  12.504 < 2e-16 ***
## as.factor(year)5  2.22721    0.15904  14.004 < 2e-16 ***
## as.factor(year)6 -0.49356    0.23250  -2.123  0.03377 *
## statewarmed:originExotic -0.10904    0.43508  -0.251  0.80210
## statewarmed:originNative 0.07068    0.41642   0.170  0.86522
## Log(theta)     -0.15837    0.09430  -1.679  0.09307 .
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -1.1839    0.8900  -1.330  0.183454
## statewarmed     0.5392    0.3964   1.360  0.173802
## originExotic   -0.6664    0.2556  -2.607  0.009139 **
## originNative   -1.9865    0.2571  -7.727  1.1e-14 ***
## as.factor(year)2  2.4473    0.8574   2.854  0.004312 **
## as.factor(year)3  3.3156    0.8726   3.800  0.000145 ***
## as.factor(year)4  3.1585    0.8639   3.656  0.000256 ***
## as.factor(year)5  2.8081    0.8626   3.255  0.001132 **
## as.factor(year)6  2.9887    0.9019   3.314  0.000920 ***
## statewarmed:originExotic -0.1857    0.4425  -0.420  0.674792
## statewarmed:originNative -0.3577    0.4303  -0.831  0.405805
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.8535
## Number of iterations in BFGS optimization: 30
## Log-likelihood: -3228 on 23 Df

```

```

# interaction between state, origin, and year
k.m12 <- zeroinfl(p_eaten ~ state * origin * year,
  dist = 'negbin',
  data = herb_kbs)
summary(k.m12)

##
## Call:
## zeroinfl(formula = p_eaten ~ state * origin * year, data = herb_kbs,
##   dist = "negbin")
##
## Pearson residuals:
##      Min       1Q   Median       3Q      Max
## -0.5746 -0.4153 -0.2949 -0.1208 11.7652
##
## Count model coefficients (negbin with log link):
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -0.8893     1.0124  -0.878  0.37972
## statearmed       0.3362     1.9130   0.176  0.86048
## originExotic    1.8451     1.0866   1.698  0.08949 .
## originNative    2.0844     1.0416   2.001  0.04539 *
## year            0.6956     0.2428   2.864  0.00418 **
## statearmed:originExotic -1.0762     2.0182  -0.533  0.59384
## statearmed:originNative -0.9461     1.9451  -0.486  0.62666
## statearmed:year    -0.1492     0.4358  -0.342  0.73200
## originExotic:year  -0.2343     0.2839  -0.825  0.40922
## originNative:year  -0.4931     0.2527  -1.951  0.05103 .
## statearmed:originExotic:year 0.3377     0.4974   0.679  0.49724
## statearmed:originNative:year 0.2362     0.4459   0.530  0.59637
## Log(theta)      -0.8249     0.1315  -6.271 3.58e-10 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)     2.1637     0.9187   2.355 0.018505 *
## statearmed       1.6470     1.8330   0.898 0.368920
## originExotic    -3.2668     0.9923  -3.292 0.000994 ***
## originNative    -4.3214     1.0533  -4.103 4.08e-05 ***
## year            -0.2324     0.2210  -1.052 0.292899
## statearmed:originExotic -1.2805     1.9195  -0.667 0.504727
## statearmed:originNative -2.5090     2.0209  -1.242 0.214402
## statearmed:year    -0.2567     0.4259  -0.603 0.546676
## originExotic:year   0.7575     0.2491   3.041 0.002358 **
## originNative:year   0.6218     0.2445   2.543 0.010992 *
## statearmed:originExotic:year 0.2740     0.4607   0.595 0.552075
## statearmed:originNative:year 0.4697     0.4579   1.026 0.305024
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.4383
## Number of iterations in BFGS optimization: 38
## Log-likelihood: -3333 on 25 Df

# state and species as fixed effects
k.m13 <- zeroinfl(p_eaten ~ state + species,

```

```

        dist = 'negbin',
        data = herb_kbs)
summary(k.m13)

##
## Call:
## zeroinfl(formula = p_eaten ~ state + species, data = herb_kbs, dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.5130 -0.4423 -0.2280 -0.1620 11.0764
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  1.85217    0.16652  11.123 < 2e-16 ***
## statewarmed -0.23848    0.11937  -1.998  0.0457 *
## speciesEugr -0.37812    0.21503  -1.758  0.0787 .
## speciesHisp  0.17181    0.28973   0.593  0.5532
## speciesPhpr  0.57539    0.30652   1.877  0.0605 .
## speciesPopr  0.77610    0.30867   2.514  0.0119 *
## speciesSoca  0.02802    0.16609   0.169  0.8660
## Log(theta)  -1.06876    0.16412  -6.512 7.41e-11 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.7171    0.6409  -2.679  0.00738 **
## statewarmed  0.2975    0.1799   1.653  0.09823 .
## speciesEugr  0.6616    0.5383   1.229  0.21905
## speciesHisp  3.1040    0.5945   5.221 1.78e-07 ***
## speciesPhpr  1.7594    0.5973   2.945  0.00323 **
## speciesPopr  3.5900    0.6027   5.956 2.58e-09 ***
## speciesSoca  0.7035    0.4821   1.459  0.14447
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.3434
## Number of iterations in BFGS optimization: 24
## Log-likelihood: -3318 on 15 Df

# state, species and year as fixed effects
k.m14 <- zeroinfl(p_eaten ~ state + species + as.factor(year),
                 dist = 'negbin',
                 data = herb_kbs)
summary(k.m14)

##
## Call:
## zeroinfl(formula = p_eaten ~ state + species + as.factor(year), data = herb_kbs,
##      dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.7568 -0.4595 -0.2437 -0.1264 24.5632
##

```

```
## Count model coefficients (negbin with log link):
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.27490    0.14310   1.921  0.0547 .
## statearmed     -0.22879    0.08973  -2.550  0.0108 *
## speciesEugr     0.30066    0.17100   1.758  0.0787 .
## speciesHispr    -0.08568    0.21913  -0.391  0.6958
## speciesPhpr     0.96249    0.23019   4.181 2.90e-05 ***
## speciesPopr     1.45768    0.23949   6.087 1.15e-09 ***
## speciesSoca     0.27379    0.12256   2.234  0.0255 *
## as.factor(year)2 1.00539    0.14799   6.794 1.09e-11 ***
## as.factor(year)3 2.11838    0.17565  12.060 < 2e-16 ***
## as.factor(year)4 2.25866    0.15781  14.313 < 2e-16 ***
## as.factor(year)5 2.18186    0.14707  14.835 < 2e-16 ***
## as.factor(year)6 -0.53881    0.23597  -2.283  0.0224 *
## Log(theta)     -0.16853    0.09807  -1.718  0.0857 .
##
## Zero-inflation model coefficients (binomial with logit link):
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -16.0545  1638.7804  -0.010  0.9922
## statearmed      0.1933    0.1357   1.425  0.1542
## speciesEugr     0.5715    0.3900   1.465  0.1429
## speciesHispr    2.8006    0.3579   7.825 5.09e-15 ***
## speciesPhpr     1.7573    0.4182   4.202 2.64e-05 ***
## speciesPopr     3.4030    0.3744   9.090 < 2e-16 ***
## speciesSoca     0.6532    0.3048   2.143  0.0321 *
## as.factor(year)2 14.6235  1638.7802   0.009  0.9929
## as.factor(year)3 15.8370  1638.7802   0.010  0.9923
## as.factor(year)4 15.4532  1638.7802   0.009  0.9925
## as.factor(year)5 14.7907  1638.7803   0.009  0.9928
## as.factor(year)6 15.2518  1638.7804   0.009  0.9926
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.8449
## Number of iterations in BFGS optimization: 50
## Log-likelihood: -3135 on 25 Df

# interaction between state and species as fixed effects, plus year
k.m15 <- zeroinfl(p_eaten ~ state * species + as.factor(year),
  dist = 'negbin',
  data = herb_kbs)
summary(k.m15)

##
## Call:
## zeroinfl(formula = p_eaten ~ state * species + as.factor(year), data = herb_kbs,
##   dist = "negbin")
##
## Pearson residuals:
##      Min       1Q   Median       3Q      Max
## -0.7484 -0.4512 -0.2436 -0.1453  24.4209
##
## Count model coefficients (negbin with log link):
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.28180    0.16239   1.735  0.0827 .
```



```

## statearmed          -0.30056    0.21826   -1.377    0.1685
## speciesEugr         0.18694    0.23344    0.801    0.4232
## speciesHisp        -0.11739    0.26062   -0.450    0.6524
## speciesPhpr         1.38715    0.35200    3.941 8.12e-05 ***
## speciesPopr         1.52144    0.30340    5.015 5.32e-07 ***
## speciesSoca         0.24827    0.15802    1.571    0.1162
## as.factor(year)2    0.98047    0.15108    6.490 8.61e-11 ***
## as.factor(year)3    2.18486    0.18141   12.044 < 2e-16 ***
## as.factor(year)4    2.25850    0.15994   14.121 < 2e-16 ***
## as.factor(year)5    2.18045    0.14810   14.723 < 2e-16 ***
## as.factor(year)6   -0.63192    0.25350   -2.493    0.0127 *
## statearmed:speciesEugr 0.28197    0.32875    0.858    0.3910
## statearmed:speciesHisp 0.15699    0.45694    0.344    0.7312
## statearmed:speciesPhpr -0.76163    0.46758   -1.629    0.1033
## statearmed:speciesPopr -0.04837    0.45134   -0.107    0.9147
## statearmed:speciesSoca 0.11182    0.24880    0.449    0.6531
## Log(theta)         -0.17952    0.10189   -1.762    0.0781 .
##
## Zero-inflation model coefficients (binomial with logit link):
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -16.0304  1675.9445  -0.010 0.992368
## statearmed     -0.1854    0.7459  -0.249 0.803662
## speciesEugr     0.2039    0.5995    0.340 0.733823
## speciesHisp     2.6398    0.4045    6.526 6.75e-11 ***
## speciesPhpr     2.0039    0.5328    3.761 0.000169 ***
## speciesPopr     3.3475    0.4313    7.761 8.45e-15 ***
## speciesSoca     0.6303    0.3477    1.812 0.069910 .
## as.factor(year)2  14.6149  1675.9445  0.009 0.993042
## as.factor(year)3  15.8883  1675.9445  0.009 0.992436
## as.factor(year)4  15.4714  1675.9445  0.009 0.992634
## as.factor(year)5  14.8115  1675.9446  0.009 0.992949
## as.factor(year)6  15.1591  1675.9447  0.009 0.992783
## statearmed:speciesEugr 0.8375    0.9352    0.895 0.370542
## statearmed:speciesHisp 0.7449    0.8458    0.881 0.378501
## statearmed:speciesPhpr -0.1405    0.8682   -0.162 0.871399
## statearmed:speciesPopr 0.4300    0.8348    0.515 0.606460
## statearmed:speciesSoca 0.3230    0.7660    0.422 0.673285
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.8357
## Number of iterations in BFGS optimization: 59
## Log-likelihood: -3131 on 35 Df

## interaction between state, species, and year - doesn't run
#m8 <- zeroinfl(p_eaten ~ state * species * year,
#               dist = 'negbin',
#               data = herb_kbs)
#summary(m8)

# likelihood ratio test
lrtest(k.m1, k.m2, k.m3, k.m4, k.m5, k.m6, k.m7, k.m8, k.m8, k.m10, k.m11, k.m12, k.m13, k.m14, k.m15)

## Likelihood ratio test
##

```

```

## Model 1: p_eaten ~ state
## Model 2: p_eaten ~ state + as.factor(year)
## Model 3: p_eaten ~ state + growth_habit
## Model 4: p_eaten ~ state + growth_habit + as.factor(year)
## Model 5: p_eaten ~ state * growth_habit
## Model 6: p_eaten ~ state * growth_habit + as.factor(year)
## Model 7: p_eaten ~ state * growth_habit * year
## Model 8: p_eaten ~ state + origin
## Model 9: p_eaten ~ state + origin
## Model 10: p_eaten ~ state * origin
## Model 11: p_eaten ~ state * origin + as.factor(year)
## Model 12: p_eaten ~ state * origin * year
## Model 13: p_eaten ~ state + species
## Model 14: p_eaten ~ state + species + as.factor(year)
## Model 15: p_eaten ~ state * species + as.factor(year)
##      #Df LogLik Df      Chisq Pr(>Chisq)
## 1      5 -3478.4
## 2     15 -3324.2 10 308.5357 < 2.2e-16 ***
## 3      9 -3340.4 -6  32.4309 1.349e-05 ***
## 4     19 -3155.4 10 369.9465 < 2.2e-16 ***
## 5     13 -3339.4 -6 367.9652 < 2.2e-16 ***
## 6     23 -3153.0 10 372.7498 < 2.2e-16 ***
## 7     25 -3266.0  2 225.8882 < 2.2e-16 ***
## 8      9 -3398.8 -16 265.6043 < 2.2e-16 ***
## 9      9 -3398.8  0  0.0000 1.000000
## 10    13 -3398.6  4  0.2785 0.991159
## 11    23 -3228.1 10 341.0731 < 2.2e-16 ***
## 12    25 -3332.6  2 209.1141 < 2.2e-16 ***
## 13    15 -3318.4 -10 28.4633 0.001521 **
## 14    25 -3135.2 10 366.3154 < 2.2e-16 ***
## 15    35 -3131.5 10  7.5151 0.676085
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

# check dispersion - chose lowest loglik model for example
E <- resid(k.m13, type = "pearson")
N <- nrow(herb_kbs)
p <- length(coef(k.m13)) + 1 # '+1' is due to theta
sum(E^2) / (N - p) # pretty close to one

```

```
## [1] 0.9637512
```

```

# pairwise comparisons
emmeans(k.m13, ~ state + species)

```

```

## state species emmean SE df asymp.LCL asymp.UCL
## ambient Cest 5.403 0.739 Inf 3.954 6.85
## warmed Cest 4.044 0.623 Inf 2.822 5.27
## ambient Eugr 3.239 0.514 Inf 2.233 4.25
## warmed Eugr 2.343 0.349 Inf 1.658 3.03
## ambient Hisp 1.513 0.396 Inf 0.737 2.29
## warmed Hisp 0.933 0.262 Inf 0.420 1.45
## ambient Phpr 5.546 1.537 Inf 2.534 8.56
## warmed Phpr 3.712 1.021 Inf 1.712 5.71
## ambient Popr 1.845 0.546 Inf 0.775 2.92
## warmed Popr 1.118 0.343 Inf 0.446 1.79

```

```
## ambient Soca      4.809 0.454 Inf      3.919      5.70
## warmed  Soca      3.469 0.333 Inf      2.816      4.12
##
## Confidence level used: 0.95
```

UMBS

```
### determining distribution ###
# first, checking for normality
hist(herb_umbs$p_eaten)
qqnorm(herb_umbs$p_eaten)
shapiro.test(herb_umbs$p_eaten)
fit <- lm(p_eaten~state, data = herb_umbs)
qqPlot(fit)
hist(herb_umbs$p_eaten[herb_umbs$state == "ambient"])
hist(herb_umbs$p_eaten[herb_umbs$state == "warmed"])
# not normal- attempting to transform data below
# log transform
herb_umbs$p_log <- log(herb_umbs$p_eaten)
hist(herb_umbs$p_log)
qqnorm(herb_umbs$p_log)
shapiro.test(herb_umbs$p_log)
```

Transformations are a no-go

Going to try a zero-inflated model due to the excess number of zeros in the data

```
# mean and var of non-zero counts
herb_umbs %>%
  dplyr::filter(p_eaten != "0") %>%
  dplyr::summarize(mean_eaten = mean(p_eaten, na.rm=T), var_eaten = var(p_eaten, na.rm=T))

## `summarise()` ungrouping output (override with `.groups` argument)

## # A tibble: 9 x 3
##   species mean_eaten var_eaten
##   <chr>      <dbl>    <dbl>
## 1 Cape        5.62      96.2
## 2 Cest       16.9      562.
## 3 Dasp       16.4      578.
## 4 Hype       27.5      622.
## 5 Poco        5.65      40.3
## 6 Popr       20.6      445.
## 7 Posp       37.1      654.
## 8 Ptaq        8.27      52.3
## 9 Ruac       22.3      606.

# variance is also > mean, so can't be poisson
# I'll try zero-inflated negative binomial due to an excess of zeros

# zero-inflated negative binomial
# state as a fixed effect
u.m1 <- zeroinfl(p_eaten ~ state,
  dist = 'negbin',
```

```

data = herb_umbs)
summary(u.m1)

##
## Call:
## zeroinfl(formula = p_eaten ~ state, data = herb_umbs, dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.4225 -0.4225 -0.3644 -0.1282  5.4643
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   2.5920     0.1030  25.172 <2e-16 ***
## statewarmed  -0.1678     0.1132  -1.482  0.138
## Log(theta)   -1.1336     0.1290  -8.785 <2e-16 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   0.03949    0.14194   0.278  0.781
## statewarmed -0.59583    0.14157  -4.209 2.57e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.3219
## Number of iterations in BFGS optimization: 11
## Log-likelihood: -4445 on 5 Df

# state and year as fixed effects
u.m2 <- zeroinfl(p_eaten ~ state + as.factor(year),
  dist = 'negbin',
  data = herb_umbs)
summary(u.m2)

##
## Call:
## zeroinfl(formula = p_eaten ~ state + as.factor(year), data = herb_umbs,
##      dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.647684 -0.465711 -0.381948  0.004589 10.296157
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -0.25914    0.21082  -1.229  0.21899
## statewarmed    0.27295    0.09317   2.930  0.00339 **
## as.factor(year)2 1.34767    0.22722   5.931 3.01e-09 ***
## as.factor(year)3 3.38400    0.22206  15.239 < 2e-16 ***
## as.factor(year)4 2.35628    0.23443  10.051 < 2e-16 ***
## as.factor(year)5 3.16140    0.23452  13.480 < 2e-16 ***
## as.factor(year)6 3.32342    0.23381  14.214 < 2e-16 ***
## Log(theta)   -0.33468    0.08254  -4.055 5.01e-05 ***
##

```

```

## Zero-inflation model coefficients (binomial with logit link):
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -10.8508    98.3571  -0.110 0.912155
## statearmed     -0.4088     0.1082  -3.777 0.000159 ***
## as.factor(year)2  10.3579    98.3571   0.105 0.916131
## as.factor(year)3  11.1118    98.3571   0.113 0.910051
## as.factor(year)4  11.9030    98.3571   0.121 0.903677
## as.factor(year)5  11.6406    98.3571   0.118 0.905790
## as.factor(year)6  11.2062    98.3571   0.114 0.909290
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.7156
## Number of iterations in BFGS optimization: 32
## Log-likelihood: -4260 on 15 Df

# state and growth habit as fixed effects
u.m3 <- zeroinfl(p_eaten ~ state + growth_habit,
                dist = 'negbin',
                data = herb_umbs)
summary(u.m3)

##
## Call:
## zeroinfl(formula = p_eaten ~ state + growth_habit, data = herb_umbs,
##          dist = "negbin")
##
## Pearson residuals:
##      Min       1Q   Median       3Q      Max
## -0.4518 -0.3987 -0.3004 -0.1529  6.0071
##
## Count model coefficients (negbin with log link):
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept)     2.58249    0.12350  20.911 < 2e-16 ***
## statearmed      -0.20063    0.11663  -1.720  0.0854 .
## growth_habitGraminoid -0.06051    0.12624  -0.479  0.6317
## Log(theta)      -1.22162    0.16327  -7.482 7.31e-14 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept)     -0.4595     0.2446  -1.879 0.060302 .
## statearmed      -0.6956     0.1909  -3.644 0.000268 ***
## growth_habitGraminoid  1.0522     0.1993   5.279 1.3e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.2948
## Number of iterations in BFGS optimization: 14
## Log-likelihood: -4415 on 7 Df

# state, growth habit, and year as fixed effects
u.m4 <- zeroinfl(p_eaten ~ state + growth_habit + as.factor(year),
                dist = 'negbin',
                data = herb_umbs)
summary(u.m4)

```

```
##
## Call:
## zeroinfl(formula = p_eaten ~ state + growth_habit + as.factor(year),
##   data = herb_umbs, dist = "negbin")
##
## Pearson residuals:
##      Min      1Q   Median      3Q      Max
## -0.66222 -0.48107 -0.33430 -0.01702 11.87829
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -0.27115    0.22055  -1.229  0.21893
## statearmed      0.29903    0.09776   3.059  0.00222 **
## growth_habitGraminoid 0.26020    0.10883   2.391  0.01680 *
## as.factor(year)2    1.03870    0.24609   4.221 2.43e-05 ***
## as.factor(year)3    3.21873    0.23991  13.417 < 2e-16 ***
## as.factor(year)4    2.28114    0.24752   9.216 < 2e-16 ***
## as.factor(year)5    3.12171    0.24711  12.633 < 2e-16 ***
## as.factor(year)6    3.29536    0.24648  13.370 < 2e-16 ***
## Log(theta)      -0.49276    0.10365  -4.754 2.00e-06 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -11.4470    117.6274  -0.097  0.92248
## statearmed     -0.3613     0.1206  -2.996  0.00274 **
## growth_habitGraminoid 1.2685     0.1648   7.697 1.39e-14 ***
## as.factor(year)2    9.5099    117.6279   0.081  0.93556
## as.factor(year)3   11.0962    117.6274   0.094  0.92484
## as.factor(year)4   11.9345    117.6274   0.101  0.91918
## as.factor(year)5   11.5427    117.6274   0.098  0.92183
## as.factor(year)6   11.7235    117.6274   0.100  0.92061
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.6109
## Number of iterations in BFGS optimization: 36
## Log-likelihood: -4219 on 17 Df

# interaction between state and growth habit as fixed effects
u.m5 <- zeroinfl(p_eaten ~ state * growth_habit,
  dist = 'negbin',
  data = herb_umbs)
summary(u.m5)

##
## Call:
## zeroinfl(formula = p_eaten ~ state * growth_habit, data = herb_umbs,
##   dist = "negbin")
##
## Pearson residuals:
##      Min      1Q   Median      3Q      Max
## -0.4419 -0.4251 -0.3161 -0.1483  6.9637
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
```

```

## (Intercept)                2.50845    0.11297   22.204 < 2e-16 ***
## statearmed                 0.04183    0.13261    0.315 0.75242
## growth_habitGraminoid      0.33677    0.19420    1.734 0.08290 .
## statearmed:growth_habitGraminoid -0.73025    0.24921   -2.930 0.00339 **
## Log(theta)                 -1.09911    0.12628   -8.704 < 2e-16 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##                               Estimate Std. Error z value Pr(>|z|)
## (Intercept)                 -0.5108    0.1959   -2.608 0.009110 **
## statearmed                  -0.1816    0.1823   -0.996 0.319271
## growth_habitGraminoid        1.4059    0.2005    7.013 2.33e-12 ***
## statearmed:growth_habitGraminoid -0.9663    0.2772   -3.486 0.000491 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.3332
## Number of iterations in BFGS optimization: 15
## Log-likelihood: -4407 on 9 Df

# interaction between state and growth habit as fixed effects, plus year
u.m6 <- zeroinfl(p_eaten ~ state * growth_habit + as.factor(year),
                dist = 'negbin',
                data = herb_umbs)
summary(u.m6)

##
## Call:
## zeroinfl(formula = p_eaten ~ state * growth_habit + as.factor(year),
##          data = herb_umbs, dist = "negbin")
##
## Pearson residuals:
##      Min       1Q   Median       3Q      Max
## -0.65834 -0.47972 -0.32606 -0.01636 11.83813
##
## Count model coefficients (negbin with log link):
##                               Estimate Std. Error z value Pr(>|z|)
## (Intercept)                 -0.270199    0.220072   -1.228 0.21953
## statearmed                   0.296797    0.106514    2.786 0.00533 **
## growth_habitGraminoid        0.254834    0.164301    1.551 0.12090
## as.factor(year)2             1.061158    0.245930    4.315 1.60e-05 ***
## as.factor(year)3             3.226367    0.239076   13.495 < 2e-16 ***
## as.factor(year)4             2.289732    0.245551    9.325 < 2e-16 ***
## as.factor(year)5             3.125807    0.245176   12.749 < 2e-16 ***
## as.factor(year)6             3.300644    0.244568   13.496 < 2e-16 ***
## statearmed:growth_habitGraminoid 0.003845    0.204963    0.019 0.98503
## Log(theta)                 -0.469147    0.102137   -4.593 4.36e-06 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##                               Estimate Std. Error z value Pr(>|z|)
## (Intercept)                -13.5562   323.7932   -0.042 0.967
## statearmed                  -0.1634    0.1515   -1.079 0.281
## growth_habitGraminoid        1.5102    0.2045    7.384 1.53e-13 ***
## as.factor(year)2            11.6530   323.7933    0.036 0.971
## as.factor(year)3            13.1255   323.7932    0.041 0.968
## as.factor(year)4            13.9538   323.7932    0.043 0.966

```

```

## as.factor(year)5          13.5468   323.7932   0.042   0.967
## as.factor(year)6          13.7481   323.7932   0.042   0.966
## statewarmed:growth_habitGraminoid  -0.5139    0.2450  -2.097   0.036 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.6255
## Number of iterations in BFGS optimization: 40
## Log-likelihood: -4217 on 19 Df

# interaction between state, growth habit, and year (year as a factor wouldn't woru - non-finite value)
u.m7 <- zeroinfl(p_eaten ~ state * growth_habit * year,
                 dist = 'negbin',
                 data = herb_umbs)
summary(u.m7)

##
## Call:
## zeroinfl(formula = p_eaten ~ state * growth_habit * year, data = herb_umbs,
##          dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.5691 -0.4850 -0.3393 -0.0292  7.7954
##
## Count model coefficients (negbin with log link):
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      1.21385    0.26728   4.541 5.59e-06 ***
## statewarmed      -0.07678    0.35475  -0.216   0.829
## growth_habitGraminoid -7.85836    0.74484 -10.550 < 2e-16 ***
## year              0.35226    0.06562   5.368 7.94e-08 ***
## statewarmed:growth_habitGraminoid  7.56913    0.87143   8.686 < 2e-16 ***
## statewarmed:year      0.02195    0.08737   0.251   0.802
## growth_habitGraminoid:year  2.98042    0.27345  10.899 < 2e-16 ***
## statewarmed:growth_habitGraminoid:year -2.85337    0.30972  -9.213 < 2e-16 ***
## Log(theta)        -0.66484    0.08994  -7.392 1.45e-13 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)     -0.56841    0.31439  -1.808   0.0706 .
## statewarmed      0.11578    0.42010   0.276   0.7828
## growth_habitGraminoid -8.71764    1.46012 -5.970 2.37e-09 ***
## year              0.10017    0.06852   1.462   0.1438
## statewarmed:growth_habitGraminoid  6.14174    1.56877   3.915 9.04e-05 ***
## statewarmed:year     -0.06972    0.09630  -0.724   0.4691
## growth_habitGraminoid:year  2.95415    0.44600   6.624 3.50e-11 ***
## statewarmed:growth_habitGraminoid:year -2.02729    0.47087  -4.305 1.67e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.5144
## Number of iterations in BFGS optimization: 23
## Log-likelihood: -4241 on 17 Df

```



```

# state and origin as fixed effects
u.m8 <- zeroinfl(p_eaten ~ state + origin,
                 dist = 'negbin',
                 data = herb_umbs)

summary(u.m8)

##
## Call:
## zeroinfl(formula = p_eaten ~ state + origin, data = herb_umbs, dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.4406 -0.4120 -0.3839 -0.1137  8.0557
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   3.4116     0.2482  13.747 < 2e-16 ***
## statearmed    -0.1406     0.1103  -1.275  0.20236
## originExotic  -0.7551     0.2509  -3.010  0.00261 **
## originNative  -1.1898     0.2641  -4.505 6.65e-06 ***
## Log(theta)    -1.0216     0.1191  -8.575 < 2e-16 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   0.3688     0.2618   1.409   0.159
## statearmed    -0.5299     0.1293  -4.097 4.19e-05 ***
## originExotic  -0.3599     0.2568  -1.401   0.161
## originNative  -0.1052     0.2716  -0.387   0.699
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.36
## Number of iterations in BFGS optimization: 16
## Log-likelihood: -4429 on 9 Df

# state, origin, and year as fixed effects
u.m9 <- zeroinfl(p_eaten ~ state + origin + as.factor(year),
                 dist = 'negbin',
                 data = herb_umbs)

summary(u.m9)

```

```

##
## Call:
## zeroinfl(formula = p_eaten ~ state + origin + as.factor(year), data = herb_umbs,
##      dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.649042 -0.471905 -0.359323 -0.008103  9.729775
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   0.13637   0.29436   0.463  0.64318
## statearmed    0.26640   0.09299   2.865  0.00417 **
## originExotic  -0.39211   0.20693  -1.895  0.05811 .

```

```

## originNative      -0.50601    0.21288  -2.377  0.01746 *
## as.factor(year)2  1.39294    0.23123   6.024 1.70e-09 ***
## as.factor(year)3  3.32660    0.22842  14.564 < 2e-16 ***
## as.factor(year)4  2.39332    0.23661  10.115 < 2e-16 ***
## as.factor(year)5  3.17364    0.23392  13.567 < 2e-16 ***
## as.factor(year)6  3.32423    0.23309  14.262 < 2e-16 ***
## Log(theta)       -0.32531    0.08300  -3.919 8.87e-05 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -13.3067   467.3575  -0.028 0.977286
## statearmed    -0.3969     0.1092  -3.636 0.000277 ***
## originExotic  -0.6512     0.2400  -2.713 0.006672 **
## originNative  -0.2539     0.2486  -1.021 0.307107
## as.factor(year)2 13.2735   467.3575   0.028 0.977342
## as.factor(year)3 13.9502   467.3575   0.030 0.976187
## as.factor(year)4 14.8811   467.3575   0.032 0.974599
## as.factor(year)5 14.6315   467.3575   0.031 0.975025
## as.factor(year)6 14.3105   467.3575   0.031 0.975573
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.7223
## Number of iterations in BFGS optimization: 40
## Log-likelihood: -4250 on 19 Df

# interaction between state and origin as fixed effects
u.m10 <- zeroinfl(p_eaten ~ state * origin,
                  dist = 'negbin',
                  data = herb_umbs)
summary(u.m10)

##
## Call:
## zeroinfl(formula = p_eaten ~ state * origin, data = herb_umbs, dist = "negbin")
##
## Pearson residuals:
##      Min      1Q   Median      3Q      Max
## -0.45402 -0.44045 -0.33874 -0.09427  7.31532
##
## Count model coefficients (negbin with log link):
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)      3.4590     0.3164  10.931 < 2e-16 ***
## statearmed       -0.1890     0.4922  -0.384  0.70098
## originExotic    -0.8762     0.3294  -2.660  0.00782 **
## originNative    -1.0417     0.3551  -2.933  0.00335 **
## statearmed:originExotic  0.1737     0.5093   0.341  0.73300
## statearmed:originNative -0.3297     0.5382  -0.613  0.54013
## Log(theta)      -1.0171     0.1202  -8.465 < 2e-16 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)     -0.6142     0.4187  -1.467  0.14237
## statearmed       1.2921     0.5153   2.507  0.01216 *
## originExotic     0.6157     0.4163   1.479  0.13922

```

```

## originNative          1.0999      0.4317    2.548  0.01085 *
## statearmed:originExotic -1.8012      0.5420   -3.323  0.00089 ***
## statearmed:originNative -2.3544      0.5871   -4.010  6.07e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.3616
## Number of iterations in BFGS optimization: 20
## Log-likelihood: -4418 on 13 Df
# interaction between state and origin as fixed effects, plus year
u.m11 <- zeroinfl(p_eaten ~ state * origin + as.factor(year),
                 dist = 'negbin',
                 data = herb_umbs)
summary(u.m11)

##
## Call:
## zeroinfl(formula = p_eaten ~ state * origin + as.factor(year), data = herb_umbs,
##          dist = "negbin")
##
## Pearson residuals:
##      Min      1Q   Median      3Q      Max
## -0.65417 -0.46593 -0.36493 -0.02109  9.47634
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.30953    0.32843   0.942  0.345968
## statearmed     -0.17994    0.37092  -0.485  0.627602
## originExotic   -0.60237    0.25997  -2.317  0.020499 *
## originNative   -0.57712    0.27344  -2.111  0.034807 *
## as.factor(year)2  1.38107    0.23181   5.958  2.56e-09 ***
## as.factor(year)3  3.31051    0.22899  14.457 < 2e-16 ***
## as.factor(year)4  2.38934    0.23691  10.085 < 2e-16 ***
## as.factor(year)5  3.15962    0.23432  13.484 < 2e-16 ***
## as.factor(year)6  3.33175    0.23351  14.268 < 2e-16 ***
## statearmed:originExotic 0.52205    0.38622   1.352  0.176470
## statearmed:originNative 0.27675    0.41098   0.673  0.500704
## Log(theta)      -0.32561    0.08377  -3.887  0.000102 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -16.2399   1319.2882  -0.012  0.990179
## statearmed      1.1347     0.4253    2.668  0.007630 **
## originExotic    0.1701     0.3413    0.498  0.618237
## originNative    0.8079     0.3537    2.284  0.022355 *
## as.factor(year)2  15.3649   1319.2882   0.012  0.990708
## as.factor(year)3  16.0297   1319.2882   0.012  0.990306
## as.factor(year)4  17.0089   1319.2882   0.013  0.989714
## as.factor(year)5  16.7121   1319.2882   0.013  0.989893
## as.factor(year)6  16.4110   1319.2882   0.012  0.990075
## statearmed:originExotic -1.5057     0.4458  -3.378  0.000731 ***
## statearmed:originNative -2.0110     0.4785  -4.202  2.64e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
##
## Theta = 0.7221
## Number of iterations in BFGS optimization: 45
## Log-likelihood: -4239 on 23 Df

## interaction between state, origin, and year - doesn't work
#u.m12 <- zeroinfl(p_eaten ~ state * origin * as.factor(year),
#                 dist = 'negbin',
#                 data = herb_umbs)
#summary(u.m12)

# state and species as fixed effects
u.m13 <- zeroinfl(p_eaten ~ state + species,
                  dist = 'negbin',
                  data = herb_umbs)
summary(u.m13)

##
## Call:
## zeroinfl(formula = p_eaten ~ state + species, data = herb_umbs, dist = "negbin")
##
## Pearson residuals:
##      Min      1Q  Median      3Q      Max
## -0.6206 -0.4564 -0.3392 -0.1233 12.7533
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  1.27936    0.30754   4.160 3.18e-05 ***
## statearmed   -0.02801    0.10706  -0.262 0.793622
## speciesCest  1.30581    0.30874   4.229 2.34e-05 ***
## speciesDasp  1.26957    0.32796   3.871 0.000108 ***
## speciesHype  1.85424    0.46366   3.999 6.36e-05 ***
## speciesPoco  0.03985    0.35454   0.112 0.910497
## speciesPopr  1.54323    0.51765   2.981 0.002871 **
## speciesPosp  2.16189    0.37543   5.759 8.49e-09 ***
## speciesPtaq  0.44828    0.34711   1.291 0.196542
## speciesRuac  1.60773    0.31775   5.060 4.20e-07 ***
## Log(theta)  -0.84553    0.09832  -8.600 < 2e-16 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   0.6326    0.3132   2.020 0.04340 *
## statearmed    -0.3529    0.1314  -2.686 0.00723 **
## speciesCest   -0.9841    0.3242  -3.036 0.00240 **
## speciesDasp   -0.2477    0.3346  -0.740 0.45904
## speciesHype    1.0087    0.4127   2.444 0.01453 *
## speciesPoco  -15.7204 1244.4038 -0.013 0.98992
## speciesPopr    2.0744    0.4333   4.788 1.69e-06 ***
## speciesPosp   -0.2279    0.3789  -0.601 0.54762
## speciesPtaq   -0.8805    0.4048  -2.175 0.02961 *
## speciesRuac   -1.0566    0.3416  -3.093 0.00198 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.4293
```

```

## Number of iterations in BFGS optimization: 42
## Log-likelihood: -4292 on 21 Df
# state, species and year as fixed effects
u.m14 <- zeroinfl(p_eaten ~ state + species + as.factor(year),
                 dist = 'negbin',
                 data = herb_umbs)
summary(u.m14)

##
## Call:
## zeroinfl(formula = p_eaten ~ state + species + as.factor(year), data = herb_umbs,
##          dist = "negbin")
##
## Pearson residuals:
##      Min      1Q   Median      3Q      Max
## -0.70260 -0.49987 -0.32878 -0.01447 11.66668
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -0.40972    0.35052  -1.169 0.242448
## statearmed     0.26343    0.09307   2.830 0.004650 **
## speciesCest    0.15545    0.28274   0.550 0.582459
## speciesDasp    0.41035    0.29147   1.408 0.159171
## speciesHype    0.38761    0.40744   0.951 0.341443
## speciesPoco    0.43451    0.31758   1.368 0.171253
## speciesPopr    0.29092    0.44567   0.653 0.513902
## speciesPosp    0.80799    0.34166   2.365 0.018035 *
## speciesPtaq   -0.01442    0.30768  -0.047 0.962609
## speciesRuac    0.49122    0.28612   1.717 0.086013 .
## as.factor(year)2 1.17523    0.24755   4.748 2.06e-06 ***
## as.factor(year)3 3.06518    0.24034  12.754 < 2e-16 ***
## as.factor(year)4 2.28318    0.24552   9.299 < 2e-16 ***
## as.factor(year)5 2.99940    0.24663  12.161 < 2e-16 ***
## as.factor(year)6 3.28438    0.23403  14.034 < 2e-16 ***
## Log(theta)     -0.33330    0.08651  -3.853 0.000117 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -14.9891  1547.7505  -0.010 0.99227
## statearmed     -0.2966    0.1191  -2.490 0.01277 *
## speciesCest    -1.4338    0.3093  -4.635 3.57e-06 ***
## speciesDasp    -0.4142    0.2989  -1.386 0.16583
## speciesHype     0.6283    0.3917   1.604 0.10871
## speciesPoco    -1.8193    0.7522  -2.419 0.01558 *
## speciesPopr     1.3788    0.4077   3.382 0.00072 ***
## speciesPosp    -0.4734    0.3713  -1.275 0.20228
## speciesPtaq    -1.0125    0.3584  -2.825 0.00473 **
## speciesRuac    -1.4319    0.3053  -4.690 2.73e-06 ***
## as.factor(year)2 15.2197  1547.7505   0.010 0.99215
## as.factor(year)3 16.0462  1547.7505   0.010 0.99173
## as.factor(year)4 16.7231  1547.7505   0.011 0.99138
## as.factor(year)5 16.1363  1547.7505   0.010 0.99168
## as.factor(year)6 16.5355  1547.7505   0.011 0.99148
## ---

```

```

## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.7166
## Number of iterations in BFGS optimization: 53
## Log-likelihood: -4154 on 31 Df

# interaction between state and species as fixed effects, plus year
u.m15 <- zeroinfl(p_eaten ~ state * species + as.factor(year),
                  dist = 'negbin',
                  data = herb_umbs)
summary(u.m15)

##
## Call:
## zeroinfl(formula = p_eaten ~ state * species + as.factor(year), data = herb_umbs,
##          dist = "negbin")
##
## Pearson residuals:
##      Min      1Q   Median      3Q      Max
## -0.71759 -0.50964 -0.31412 -0.05114 10.77704
##
## Count model coefficients (negbin with log link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -0.56238    0.46370  -1.213 0.225209
## statearmed      0.49483    0.52624   0.940 0.347058
## speciesCest     0.23959    0.42534   0.563 0.573242
## speciesDasp     0.67525    0.45045   1.499 0.133859
## speciesHype    -0.02343    0.56353  -0.042 0.966841
## speciesPoco    -0.04919    0.67903  -0.072 0.942252
## speciesPopr     0.46320    1.27083   0.364 0.715497
## speciesPosp     1.17204    0.48351   2.424 0.015348 *
## speciesPtaq     0.38266    0.47107   0.812 0.416613
## speciesRuac     0.77307    0.43323   1.784 0.074354 .
## as.factor(year)2 1.19443    0.24644   4.847 1.26e-06 ***
## as.factor(year)3 3.01325    0.24087  12.510 < 2e-16 ***
## as.factor(year)4 2.27642    0.24396   9.331 < 2e-16 ***
## as.factor(year)5 2.96236    0.24562  12.061 < 2e-16 ***
## as.factor(year)6 3.30472    0.23268  14.203 < 2e-16 ***
## statearmed:speciesCest -0.09171    0.54453  -0.168 0.866259
## statearmed:speciesDasp -0.38308    0.57492  -0.666 0.505211
## statearmed:speciesHype  0.79783    0.77615   1.028 0.303981
## statearmed:speciesPoco  0.45497    0.77816   0.585 0.558771
## statearmed:speciesPopr -0.20828    1.36477  -0.153 0.878705
## statearmed:speciesPosp -0.67450    0.64178  -1.051 0.293267
## statearmed:speciesPtaq -0.63943    0.60653  -1.054 0.291767
## statearmed:speciesRuac -0.41412    0.55624  -0.744 0.456576
## Log(theta)      -0.30399    0.08523  -3.567 0.000361 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -13.70447  1126.42094  -0.012 0.990293
## statearmed     -2.54202    0.67898  -3.744 0.000181 ***
## speciesCest    -1.97446    0.40609  -4.862 1.16e-06 ***
## speciesDasp    -0.79081    0.41200  -1.919 0.054930 .
## speciesHype    -0.15737    0.54007  -0.291 0.770755

```

```

## speciesPoco          -2.58422    2.17122   -1.190  0.233961
## speciesPopr           2.48657    1.07616    2.311  0.020855 *
## speciesPosp          -1.90334    0.50200   -3.792  0.000150 ***
## speciesPtaq          -2.04852    0.51673   -3.964  7.36e-05 ***
## speciesRuac          -2.43632    0.41879   -5.818  5.97e-09 ***
## as.factor(year)2      14.64934  1126.42090    0.013  0.989624
## as.factor(year)3      15.40399  1126.42088    0.014  0.989089
## as.factor(year)4      16.12980  1126.42088    0.014  0.988575
## as.factor(year)5      15.44148  1126.42088    0.014  0.989063
## as.factor(year)6      15.89701  1126.42088    0.014  0.988740
## statearmed:speciesCest  2.04479    0.71433    2.863  0.004203 **
## statearmed:speciesDasp  1.68122    0.73296    2.294  0.021805 *
## statearmed:speciesHype  2.47526    0.88283    2.804  0.005051 **
## statearmed:speciesPoco  2.24483    2.35794    0.952  0.341083
## statearmed:speciesPopr -0.04895    1.26292   -0.039  0.969080
## statearmed:speciesPosp  3.67473    0.80065    4.590  4.44e-06 ***
## statearmed:speciesPtaq  2.83563    0.80840    3.508  0.000452 ***
## statearmed:speciesRuac  2.90264    0.72344    4.012  6.01e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Theta = 0.7379
## Number of iterations in BFGS optimization: 69
## Log-likelihood: -4123 on 47 Df

## interaction between state, species, and year - doesn't run
#m8 <- zeroinfl(p_eaten ~ state * species * year,
#              dist = 'negbin',
#              data = herb_umbs)
#summary(m8)

# likelihood ratio test
lrtest(u.m1, u.m2, u.m3, u.m4, u.m5, u.m6, u.m7, u.m8, u.m8, u.m10, u.m11, u.m13, u.m14, u.m15)

## Likelihood ratio test
##
## Model 1: p_eaten ~ state
## Model 2: p_eaten ~ state + as.factor(year)
## Model 3: p_eaten ~ state + growth_habit
## Model 4: p_eaten ~ state + growth_habit + as.factor(year)
## Model 5: p_eaten ~ state * growth_habit
## Model 6: p_eaten ~ state * growth_habit + as.factor(year)
## Model 7: p_eaten ~ state * growth_habit * year
## Model 8: p_eaten ~ state + origin
## Model 9: p_eaten ~ state + origin
## Model 10: p_eaten ~ state * origin
## Model 11: p_eaten ~ state * origin + as.factor(year)
## Model 12: p_eaten ~ state + species
## Model 13: p_eaten ~ state + species + as.factor(year)
## Model 14: p_eaten ~ state * species + as.factor(year)
##      #Df LogLik Df   Chisq Pr(>Chisq)
## 1      5 -4445.5
## 2     15 -4260.0 10 370.954 < 2.2e-16 ***
## 3      7 -4415.4 -8 310.842 < 2.2e-16 ***
## 4     17 -4218.9 10 393.026 < 2.2e-16 ***

```

```

## 5      9 -4407.2 -8 376.532 < 2.2e-16 ***
## 6     19 -4216.7 10 381.040 < 2.2e-16 ***
## 7     17 -4240.7 -2  48.131 3.536e-11 ***
## 8      9 -4428.5 -8 375.526 < 2.2e-16 ***
## 9      9 -4428.5  0  0.000 1.0000000
## 10    13 -4417.6  4  21.726 0.0002272 ***
## 11    23 -4238.8 10 357.601 < 2.2e-16 ***
## 12    21 -4292.4 -2 107.162 < 2.2e-16 ***
## 13    31 -4153.7 10 277.526 < 2.2e-16 ***
## 14    47 -4123.0 16  61.233 3.239e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# check dispersion - chose lowest loglik model for example
E <- resid(u.m6, type = "pearson")
N <- nrow(herb_umbs)
p <- length(coef(u.m6)) + 1 # '+1' is due to theta
sum(E^2) / (N - p) # pretty close to one

## [1] 1.037604

# pairwise comparisons
emmeans(u.m13, ~ state + species)

##   state  species emmean    SE df asymp.LCL asymp.UCL
##   ambient Cape      1.25 0.372 Inf      0.5174      1.98
##   warmed  Cape      1.50 0.443 Inf      0.6357      2.37
##   ambient Cest      7.79 0.825 Inf      6.1707      9.40
##   warmed  Cest      8.63 0.789 Inf      7.0864     10.18
##   ambient Dasp      5.18 0.794 Inf      3.6255      6.74
##   warmed  Dasp      6.12 0.934 Inf      4.2910      7.95
##   ambient Hype      3.73 1.504 Inf      0.7777      6.67
##   warmed  Hype      4.82 1.892 Inf      1.1164      8.53
##   ambient Poco      3.74 0.786 Inf      2.2009      5.28
##   warmed  Poco      3.64 0.674 Inf      2.3166      4.96
##   ambient Popr      1.05 0.508 Inf      0.0561      2.05
##   warmed  Popr      1.42 0.667 Inf      0.1110      2.73
##   ambient Posp     12.50 3.131 Inf      6.3591     18.63
##   warmed  Posp     14.79 3.649 Inf      7.6359     21.94
##   ambient Ptaq      3.16 0.590 Inf      2.0035      4.32
##   warmed  Ptaq      3.53 0.620 Inf      2.3197      4.75
##   ambient Ruac     10.84 1.350 Inf      8.1980     13.49
##   warmed  Ruac     11.95 1.458 Inf      9.0938     14.81
##
## Confidence level used: 0.95

```