Mixed Effects Models

Lecture 06.2: Generalized Mixed Effects Models

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Module: Linear, Nonlinear, and Mixed Effects Models

Readings

Required for class:

► NA

Optional:

- ▶ Bolker et al. (2009) Generalized linear mixed models: apractical guide for ecology and evolution. *TREE*.
- ▶ glmer in lme4

Generalized Mixed Effects Models

When our data are not normal and we have fixed and random effects, we need to used *Generalized* Mixed Effects Models. The family specifications follows the same form as linear models. Some common types of generalized mixed effects models include:

- 1. Poisson for count data (discrete) 2. Binomial for 0/1 data
- 3. Inverse Gaussian for skewed data (continuous)

Model Statements with glmer()

This is part of the lme4 library.

Data

Let's examine this paper that explores ungulate behavior in response to refuges in urban landscapes in India.



RESEARCH ARTICLE

Wild Ungulate Decision-Making and the Role of Tiny Refuges in Human-Dominated Landscapes

Yarlagadda Chaitanya Krishna^{1,2,3}*, Ajith Kumar³, Kavita Isvaran¹

Blackbuck Data

Blackbuck are group-living antelope found on the Indian subcontinent. Krishna et al. hypothesize that in urban settings, blackbuck will be more risk-averse when food availability is high and vice versa.



Blackbuck Data

Does Blackbuck occurrence (Blackbuck) depend on how open a landscape is (Openness) and it's resource availability as measured by plant biomass (scaled_wt)? The points at which the observations were taken (Pointcode) should be the random intercept. We should use family = poisson here because we have count data.

Blackbuck Results

To get the model results, we use summary() for glmer.

```
summary(model_glm)
## Generalized linear mixed model fit by maximum likelihood (Laplace
    Approximation) [glmerMod]
## Family: poisson (log)
## Formula: Blackbuck ~ Openness + scaled wt + (1 | Pointcode)
     Data: buck
##
##
##
       ATC
                BIC logLik deviance df.resid
              877 5 -427 4 854 8
##
     862.8
                                          286
##
## Scaled residuals:
      Min 10 Median
                                     Max
## -1.4889 -0.7483 -0.4999 0.4797 3.7342
##
## Random effects:
## Groups
             Name
                    Variance Std.Dev.
## Pointcode (Intercept) 0.9435 0.9714
## Number of obs: 290, groups: Pointcode, 38
##
## Fixed effects:
##
               Estimate Std. Error z value Pr(>|z|)
## (Intercept) 0.590725 0.449144 1.315 0.188435
## Openness -0.004648 0.004598 -1.011 0.312020
## scaled_wt -0.044919 0.012680 -3.543 0.000396 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
            (Intr) Opnnss
## Openness -0.905
```

Variance Components

Again, we can determine the contribution that the fixed and random effects have on the variance explained.

```
r.squaredGLMM(model_glm)

## R2m R2c

## delta 0.04454406 0.5909729

## lognormal 0.04947089 0.6563379

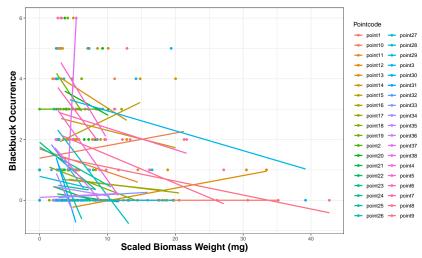
## trigamma 0.03788309 0.5026009
```

For GLMM's, you get multiple types of \mathbb{R}^2 values depending on the method used (e.g. trigamma or lognormal) for the calculations for the binomial family. They should be similar so just state which method you report.

Again, the fixed effects explain a small amount of the variance $(\sim 4\%)$ vs the fixed and random together $(\sim 60\%)$.

Blackbuck Results

To visualize these results, let's plot the blackbuck occurrence against the scaled biomass data and demonstrate the random effects.



Nonlinear Mixed Effects Models

We are not going to go over these, but I wanted to make you all aware that there are non-linear mixed effects models if your data follows a different functional form that is not linear (e.g. quadratic, parabolic, etc).

For more information, check out the nlmer() function in lme4.