

Generalized Linear Mixed Models

Quantitative Analysis of Vertebrate Populations

Generalize Linear Model

Example with a Poisson distribution for count data. **Why do we use Poisson for counts?**

1. Distribution: $C_i \sim \text{Poisson}(\lambda_i)$
2. Link Function: $\log(\lambda_i)$
3. Linear Predictor: $\log(\lambda_i) = \alpha + \beta X_i$

Generalized Linear Mixed Model

Add submodels for the random effect parameters.

1. Distribution: $C_i \sim \text{Poisson}(\lambda_i)$
2. Link Function: $\log(\lambda_i)$
3. Linear Predictor: $\log(\lambda_i) = \alpha_{j(i)} + \beta X_i$
4. Submodel for random effects parameters

$$\alpha_j \sim \mathcal{N}(\mu_\alpha, \sigma_\alpha^2)$$

Random intercept and slope model

1. Distribution: $C_i \sim \text{Poisson}(\lambda_i)$
2. Link Function: $\log(\lambda_i)$
3. Linear Predictor: $\log(\lambda_i) = \alpha_{j(i)} + \beta_{j(i)} X_i$
4. Submodel for random effects parameters

$$\alpha_j \sim \mathcal{N}(\mu_\alpha, \sigma_\alpha^2)$$

$$\beta_j \sim \mathcal{N}(\mu_\beta, \sigma_\beta^2)$$

Data generation (From Kery 2010)

Random slope and intercepts without correlation

```
n.groups <- 16
n.years <- 30
n <- n.groups * n.years
pop <- gl(n = n.groups, k = n.years)
```

Standardize the year covariate to range from zero to one.

```
original.year <- rep(1:n.years, n.groups)
year <- (original.year-1)/29
```

Build design matrix without intercept

```
Xmat <- model.matrix(~pop*year-1-year)
print(Xmat[1:21,], dig = 2)      # Print top 21 rows
```

```
##      pop1 pop2 pop3 pop4 pop5 pop6 pop7 pop8 pop9 pop10 pop11 pop12 pop13
## 1      1  0  0  0  0  0  0  0  0  0  0  0  0
## 2      1  0  0  0  0  0  0  0  0  0  0  0  0
## 3      1  0  0  0  0  0  0  0  0  0  0  0  0
## 4      1  0  0  0  0  0  0  0  0  0  0  0  0
## 5      1  0  0  0  0  0  0  0  0  0  0  0  0
## 6      1  0  0  0  0  0  0  0  0  0  0  0  0
## 7      1  0  0  0  0  0  0  0  0  0  0  0  0
## 8      1  0  0  0  0  0  0  0  0  0  0  0  0
## 9      1  0  0  0  0  0  0  0  0  0  0  0  0
## 10     1  0  0  0  0  0  0  0  0  0  0  0  0
## 11     1  0  0  0  0  0  0  0  0  0  0  0  0
## 12     1  0  0  0  0  0  0  0  0  0  0  0  0
## 13     1  0  0  0  0  0  0  0  0  0  0  0  0
## 14     1  0  0  0  0  0  0  0  0  0  0  0  0
## 15     1  0  0  0  0  0  0  0  0  0  0  0  0
## 16     1  0  0  0  0  0  0  0  0  0  0  0  0
## 17     1  0  0  0  0  0  0  0  0  0  0  0  0
## 18     1  0  0  0  0  0  0  0  0  0  0  0  0
## 19     1  0  0  0  0  0  0  0  0  0  0  0  0
## 20     1  0  0  0  0  0  0  0  0  0  0  0  0
## 21     1  0  0  0  0  0  0  0  0  0  0  0  0
##      pop14 pop15 pop16 pop1:year pop2:year pop3:year pop4:year pop5:year
## 1      0  0  0  0.000      0      0      0      0
## 2      0  0  0  0.034      0      0      0      0
## 3      0  0  0  0.069      0      0      0      0
## 4      0  0  0  0.103      0      0      0      0
## 5      0  0  0  0.138      0      0      0      0
## 6      0  0  0  0.172      0      0      0      0
## 7      0  0  0  0.207      0      0      0      0
## 8      0  0  0  0.241      0      0      0      0
## 9      0  0  0  0.276      0      0      0      0
## 10     0  0  0  0.310      0      0      0      0
## 11     0  0  0  0.345      0      0      0      0
## 12     0  0  0  0.379      0      0      0      0
## 13     0  0  0  0.414      0      0      0      0
## 14     0  0  0  0.448      0      0      0      0
## 15     0  0  0  0.483      0      0      0      0
## 16     0  0  0  0.517      0      0      0      0
## 17     0  0  0  0.552      0      0      0      0
## 18     0  0  0  0.586      0      0      0      0
## 19     0  0  0  0.621      0      0      0      0
## 20     0  0  0  0.655      0      0      0      0
## 21     0  0  0  0.690      0      0      0      0
##      pop6:year pop7:year pop8:year pop9:year pop10:year pop11:year
## 1      0      0      0      0      0      0
## 2      0      0      0      0      0      0
## 3      0      0      0      0      0      0
## 4      0      0      0      0      0      0
```

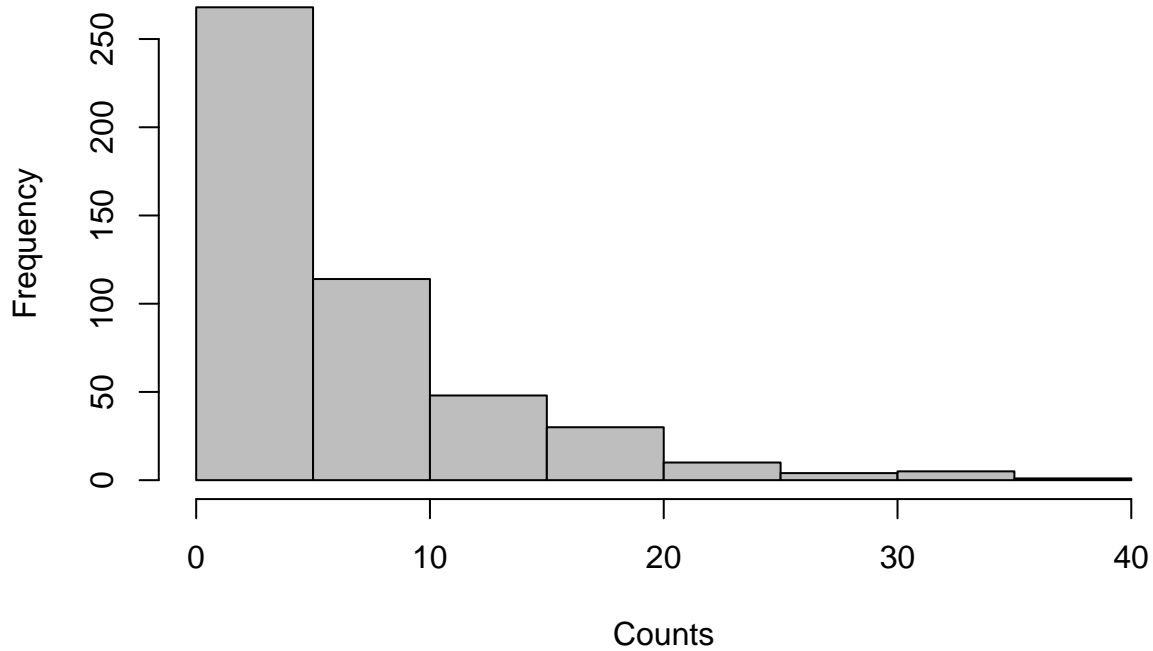
```
## 5      0      0      0      0      0      0
## 6      0      0      0      0      0      0
## 7      0      0      0      0      0      0
## 8      0      0      0      0      0      0
## 9      0      0      0      0      0      0
## 10     0      0      0      0      0      0
## 11     0      0      0      0      0      0
## 12     0      0      0      0      0      0
## 13     0      0      0      0      0      0
## 14     0      0      0      0      0      0
## 15     0      0      0      0      0      0
## 16     0      0      0      0      0      0
## 17     0      0      0      0      0      0
## 18     0      0      0      0      0      0
## 19     0      0      0      0      0      0
## 20     0      0      0      0      0      0
## 21     0      0      0      0      0      0
##      pop12:year pop13:year pop14:year pop15:year pop16:year
## 1      0      0      0      0      0
## 2      0      0      0      0      0
## 3      0      0      0      0      0
## 4      0      0      0      0      0
## 5      0      0      0      0      0
## 6      0      0      0      0      0
## 7      0      0      0      0      0
## 8      0      0      0      0      0
## 9      0      0      0      0      0
## 10     0      0      0      0      0
## 11     0      0      0      0      0
## 12     0      0      0      0      0
## 13     0      0      0      0      0
## 14     0      0      0      0      0
## 15     0      0      0      0      0
## 16     0      0      0      0      0
## 17     0      0      0      0      0
## 18     0      0      0      0      0
## 19     0      0      0      0      0
## 20     0      0      0      0      0
## 21     0      0      0      0      0
```

Draw random intercept and slopes from normal distributions

```
intercept.mean <- 3           # Choose values for the hyperparams
intercept.sd <- 1
slope.mean <- -2
slope.sd <- 0.6
intercept.effects<-rnorm(n = n.groups, mean = intercept.mean, sd = intercept.sd)
slope.effects <- rnorm(n = n.groups, mean = slope.mean, sd = slope.sd)
all.effects <- c(intercept.effects, slope.effects) # Put them all together
```

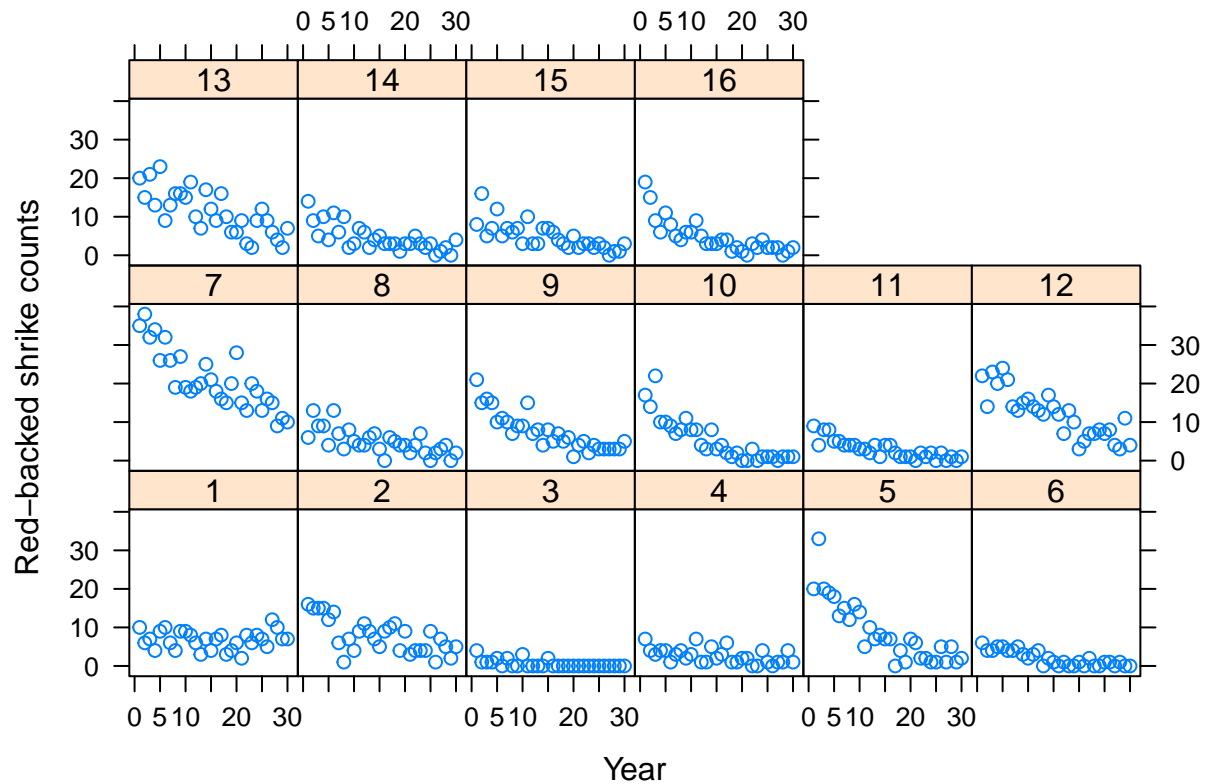
Assemble and inspect results

```
lin.pred <- Xmat[,] %*% all.effects # Value of lin.predictor  
C <- rpois(n = n, lambda = exp(lin.pred)) # Exponentiate and add Poisson noise  
hist(C, col = "grey", xlab = "Counts", main = "") # Inspect what we've created
```



Assemble and inspect results

```
library("lattice")  
xyplot(C ~ original.year | pop, ylab = "Red-backed shrike counts", xlab = "Year")
```



Analysis under a random-coefficients model

```
library('lme4')
glmm.fit <- glmer(C ~ year + (1 | pop) + (0 + year | pop), family = poisson)
summary(glmm.fit) # Inspect results
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: poisson ( log )
## Formula: C ~ year + (1 | pop) + (0 + year | pop)
##
##      AIC      BIC    logLik deviance df.resid
##  2132.5   2149.2  -1062.2   2124.5     476
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.8614 -0.6851 -0.1275  0.5581  2.7093
##
## Random effects:
##  Groups Name      Variance Std.Dev.
##  pop      (Intercept) 0.4560  0.6753
##  pop.1    year        0.9218  0.9601
## Number of obs: 480, groups: pop, 16
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   2.4356     0.1729  14.087 < 2e-16 ***
```

```
## year          -2.0116      0.2582  -7.792 6.62e-15 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##      (Intr)
## year -0.059
```

Examine Pearson residuals for patterning

```
plot(glm.fit)
```

