

# New models for reptile and amphibian removal data

Eleni Matechou, Rachel McCrea, Byron Morgan, Richard Griffiths, David Sewell and Brett Lewis

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# Result of fitting geometric model to GCN data

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Introduction and data

Complexity: covariates & clearances

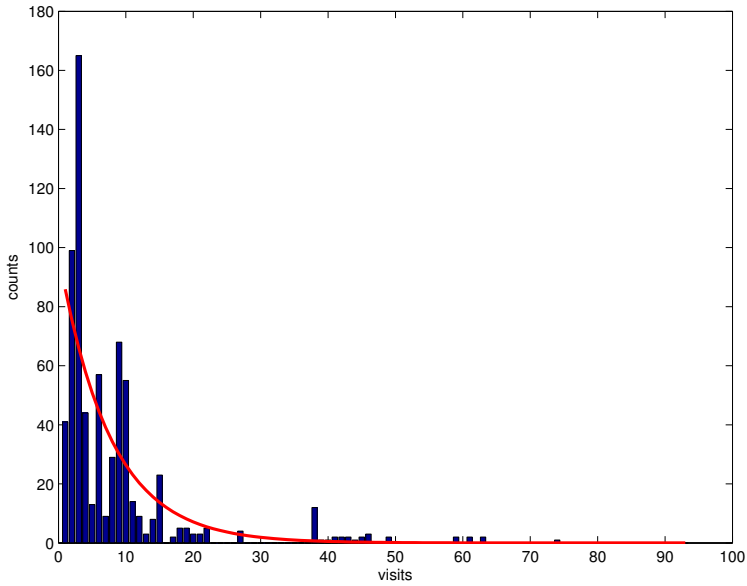
Stop-over model

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# and then adding over dispersion

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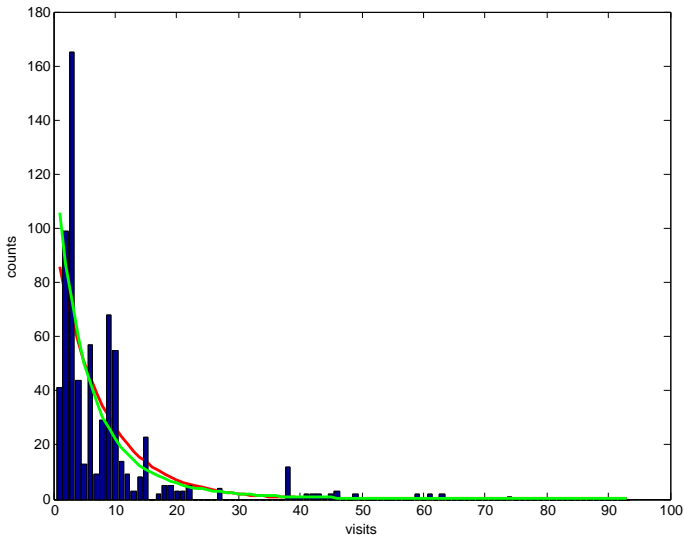
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# Geometric removal model with min air temperature

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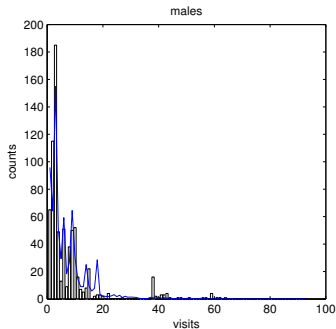
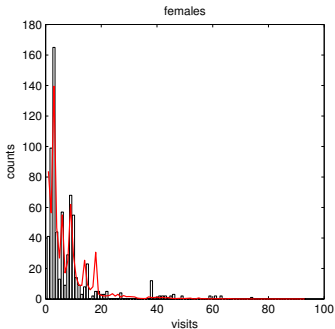
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# Timed clearance data: mainly slow worms

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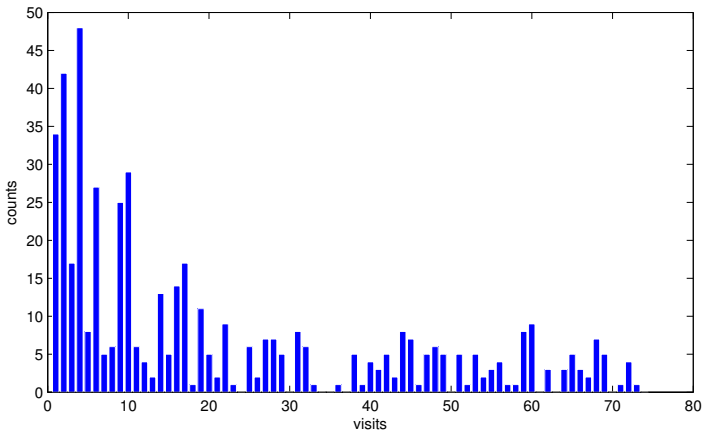
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# Model for timed clearances

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- For the **timed clearance** data, at fixed times parts of the study area were cleared.
- This was modelled by assuming a global population of animals of size  $N$ , to be estimated.
- Fractions of this number were assumed to be available for capture during each time interval. Cf **Stop-over modelling**.
- Using maximum likelihood, the fractions were estimated as: 0.45, 0.10, 0.17, 0.28, 0.00.
- The estimated number of animals not observed was  $\hat{N}_0 = 51$ .

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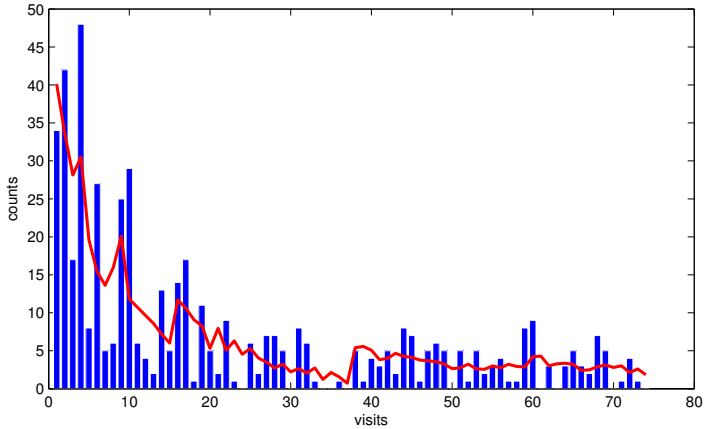
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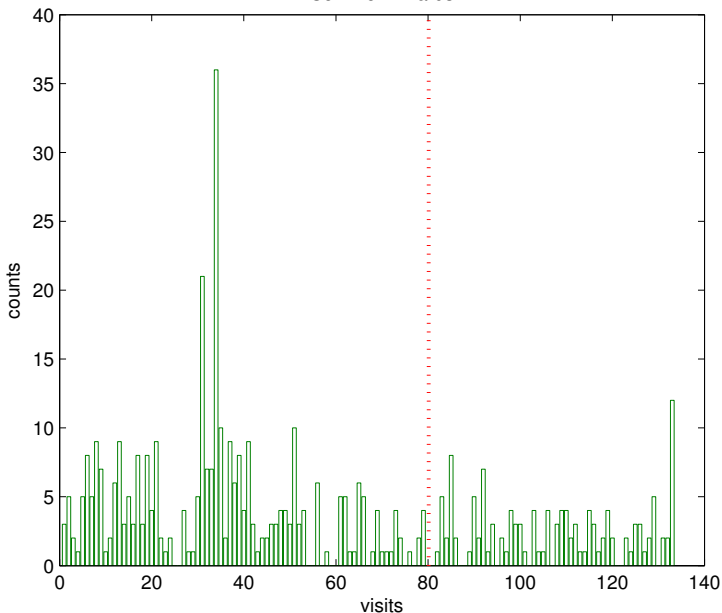
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## Common Lizards





## Stop-over model parameters

- $N$ : population size;  $M$ : number of arrival groups.
- $w_m$ ,  $\mu_m$  and  $\sigma_m$ ,  $m = 1, \dots, M$ : The population fractions, mean arrival times and standard deviations of arrival times of the  $M$  arrival groups,  $\sum_{m=1}^M w_m = 1$ . The population fraction that arrived between occasions  $b - 1$  and  $b$  is denoted by  $\beta_{b-1}$ . In terms of the mixture components,

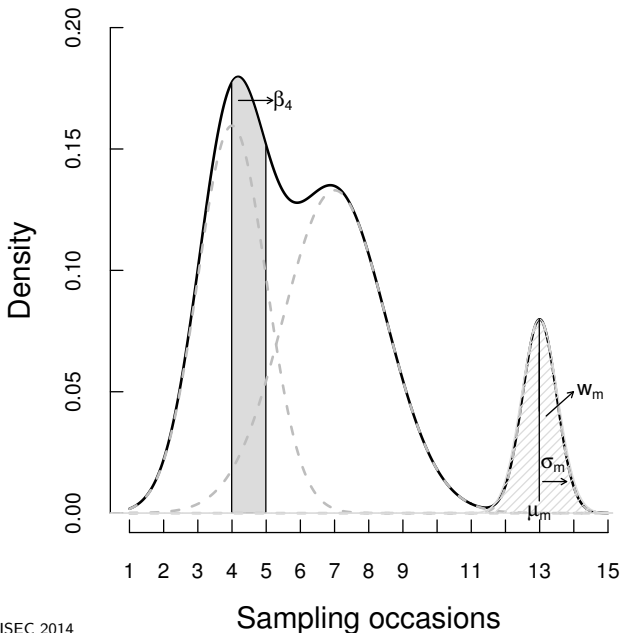
$$\beta_{b-1} = \sum_{m=1}^M w_m \{F_m(b) - F_m(b-1)\}, \quad b = 2, \dots, T-1$$

where  $F_m(b) = P(X \leq b)$  when  $X \sim N(\mu_m, \sigma_m^2)$ . The first and last intervals are open-ended with

$$\beta_0 = \sum_{m=1}^M w_m F_m(1) \text{ and}$$

$\beta_{T-1} = 1 - \sum_{m=1}^M w_m F_m(T-1)$ ,  $\forall m$ , ensuring that the entry parameters sum to 1 i.e.  $\sum_{b=1}^T \beta_{b-1} = 1$ .

## Stop-over model: $M = 3$



# Forming the likelihood

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If an individual belongs to category  $t$ ,  $t = 1, \dots, T$ , then it was removed on sampling occasion  $t$ . The unknown number of individuals that were never detected and therefore never removed belong to category  $T + 1$ .

The **probability of belonging to category**  $t$ ,  $\gamma_t$ , is:

$$\gamma_t = \begin{cases} \sum_{b=1}^t \left[ \beta_{b-1} \left\{ \prod_{k=b}^{t-1} (1 - p_k) \right\} \right] p_t, & t = 1, \dots, T \\ 1 - \sum_{t=1}^T \gamma_t = \sum_{b=1}^T \left[ \beta_{b-1} \prod_{k=b}^T (1 - p_k) \right], & t = T + 1 \end{cases},$$

The likelihood is **multinomial** with  $T + 1$  cells,  $\gamma_t$ ,  $t = 1, \dots, T + 1$  probabilities and  $n_t$ ,  $t = 1, \dots, T + 1$  frequencies.

# Stop-over likelihood

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The **parameters** are given by

$$\boldsymbol{\theta} = (M, N, (w_m, \mu_m, \sigma_m)_{m=1, \dots, M}, (p_t)_{t=1, \dots, T})$$

and the **likelihood** is:

$$p(\mathbf{y}|\boldsymbol{\theta}) = \frac{N!}{\left(\prod_{t=1}^T n_t!\right) (N-D)!} \left\{ \prod_{t=1}^T \gamma_t^{n_t} \right\} \gamma_{T+1}^{N-D},$$

where  $D = \sum_{t=1}^T n_t$ . We assume constant capture in the applications.

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# Stop-over, RJMCMC, model averaged, GCN

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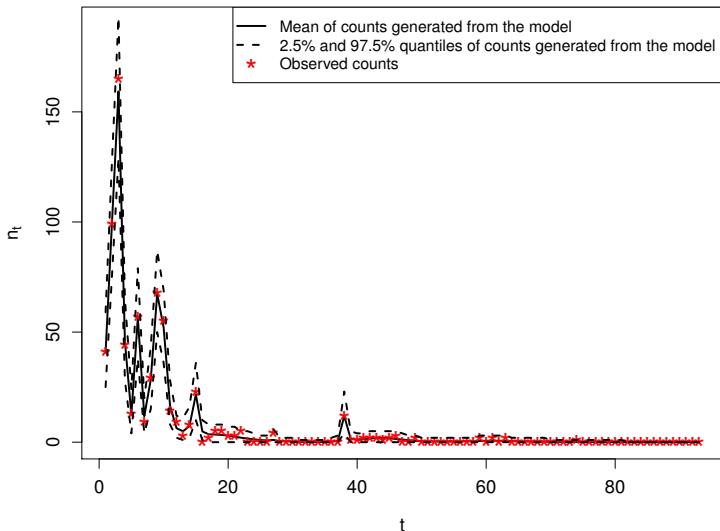
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# Stop-over, RJMCMC, model averaged, lizard

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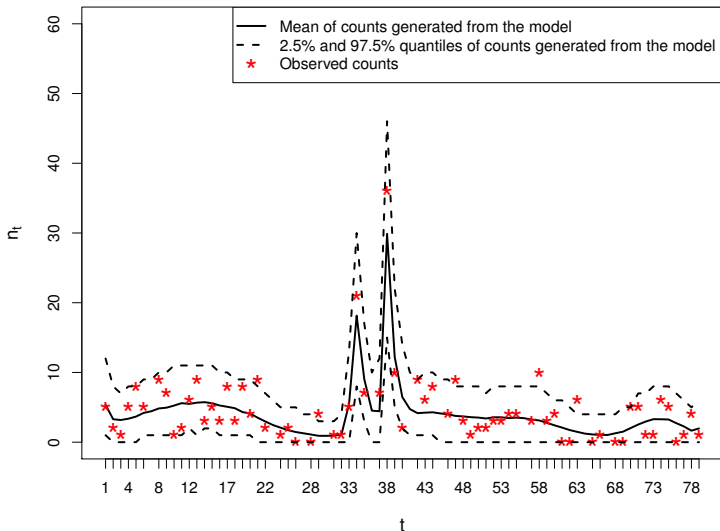
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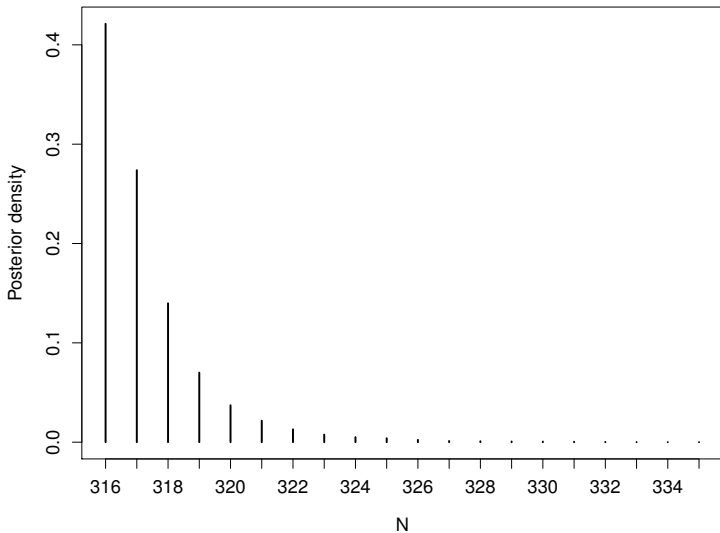
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## Posterior distribution for $N$ for lizard



# References

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- 5 Zippin (1958) *J. Wildlife Management*, **22**, 82–90.

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# Covariates

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- In practice we can expect the **capture probability to vary**
- for fecundability, conception might **vary with age and parity**
- for animals capture might vary with **temperature**. For GCN, **minimum air temperature** was used
- We may have a **logistic** transformation, for a covariate  $w$ :

$$p = \frac{1}{1 + \exp(\alpha + \beta w)}.$$

- It is often necessary to **choose** the best covariate(s) from a relevant set.
- **Overdispersion** may also be included, eg., using a **beta-geometric** distribution.

# Modelling

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- The **simple removal model** dates from papers by Moran (1951) and Zippin (1958)
- It assumes a **constant capture probability**,  $p$ .
- The other parameter is the desired population size,  $N$ .
- The same model applies to **fecundability** data, which record months to **conception** for human couples.
- **100** Smokers: 29 16 17 4 3 9 4 5 1 1 1 3 7;  $\hat{N} = 95$ .
- **486** Non-smokers: 198 107 55 38 18 22 7 9 5 3 6 6 12;  $\hat{N} = 476$ .

## Notation and likelihood for geometric model

- $N$ : initial population size
- $n_k$ : size of the  $k^{\text{th}}$  sample removed from the population,  $k = 1, 2, \dots, T$
- $x_k = \sum_{j=1}^{k-1} n_j$ ,  $k = 2, 3, \dots, T + 1$ ;  $x_1 = 0$ .

For example:  $\mathbf{n} = (65, 115, 185, \dots)$

$\mathbf{x} = (0, 65, 180, 365, \dots)$ .

We then form the likelihood:

$$L(N, p; \mathbf{n}) = \frac{N!}{(\prod_{k=1}^T n_k!)(N - x_{T+1})!} \left[ \prod_{k=1}^T \{p(1-p)^{k-1}\}^{n_k} \right] (1-p)^{T(N-x_{T+1})}$$

# Maximum-likelihood estimates

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The likelihood simplifies to give

$$L(N, p; \mathbf{n}) = \frac{N!}{(\prod_{k=1}^T n_k!)(N - x_{T+1})!} p^{x_{T+1}} (1-p)^{TN - \sum_{k=1}^{T+1} x_k}$$

**Maximum-likelihood estimates** of the two parameters are given by the solutions to the equations:

$$\hat{N} = \frac{x_{T+1}}{1 - (1 - \hat{p})^T}$$

$$\frac{\hat{p}}{1 - \hat{p}} - \frac{T(1 - \hat{p})^T}{1 - (1 - \hat{p})^T} = \frac{\sum_{k=1}^T (k-1)n_k}{x_{T+1}}$$

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