

Stat 534: formulae referenced in lecture, week 8b:  
Occupancy, N-mixture and Spatial models

Occupancy modeling

- Notation:
  - $X_i$ : characteristics of site  $i$ , may be a vector
  - $Y_i$ : species is seen at site  $i$ ,  
0 = not seen, 1 = seen
  - $Z_i$ : species occurs at site  $i$ ,  
0 = doesn't occur there, 1 = does

- The model:

$$\begin{aligned}Z_i &\sim \text{Bern}(\pi_i) \\ \pi_i &= f(\beta, X_i) \\ Y_i | Z_i &\sim \text{Bern}(Z_i \times g(\alpha, X_i))\end{aligned}$$

N-mixture models:

- Estimate  $N$  and  $p$  without marking animals
- Notation:
  - $N_i$ : # individuals at site  $i$
  - $S$  sites, each sampled  $T$  times
  - $n_{ij}$ : # individuals in  $j$  sample at site  $i$
  - $p$ : constant, or function of known covariates
- The model (Royle 2004; Kéry, Royle and Schmid, 2005)

$$\begin{aligned}n_{ij} &\sim \text{Bin}(N_i, p) \\ N_i &\sim f(\theta)\end{aligned}$$

- log-likelihood:

$$\ln L(\theta, p | \{n_{ij}\}) = \prod_{i=1}^S \left\{ \sum_{N_i=k_i}^{\infty} \left( \prod_{j=1}^T \text{Bin}(n_{ij} | N_i, p) \right) f(N_i | \theta) \right\}$$

- Why multiple  $T > 1$  times are needed:
  - $T = 1$ :  $N_i \sim \text{Pois}(\lambda_i)$ ,  $n_i | N_i \sim \text{Bin}(N_i, p) \Rightarrow n_i \sim \text{Pois}(\lambda_i p)$
  - Can't estimate  $\lambda_i$  or  $N_i$ , only relative abundance:  $\lambda_i / \lambda_j$
  - $T = 2$ :  $n_{ij} \sim \text{Pois}(\lambda_i p)$ .
  - $n_{i1}$  and  $n_{i2}$  are correlated
  - That correlation  $\Rightarrow \hat{\lambda}_i$
  - But inference from correlations is fragile

Distance sampling: just pictures in hand-written notes

Spatial capture-recapture:

- Notation:
  - $K$ : # capture occasions,  $J$ : # traps,  $N$ : # individuals
  - $Y_{ij}$ : # times animal  $i$  caught in trap  $j \sim \text{Bin}(K, P_{ij})$
  - $p_{ij}$ :  $P[\text{animal } i \text{ caught in trap } j]$
  - $S_i$ : activity center for animal  $i$
  - $X_j$ : location of trap  $j$
  - $\alpha$ : parameters of home range / activity distributions
- The model, omitting some details
  - $p_{ij}$  depends on home-range “shape” and distance between  $S_i$  and  $X_j$
  - Gaussian home ranges:  

$$p_{ij} = \alpha_0 \exp(-\alpha_1 \|S_i - X_j\|)$$
    - \*  $\alpha_0$ : capture probability when trap at activity center
    - \*  $\alpha_1$ : describes “spread” of home range
  - Big assumption: home range is isotropic only distance matters, not direction
  - For animal  $i$ :

$$f(Y_{ij} | S_i) = \prod_{traps} \text{Bin}(k, p_{ij})$$

– Want  $f(Y_{ij})$ :

$$f(Y_{ij}) = \int f(Y_{ij} | S_i) f(S_i) dS_i$$

– requires a distribution for activity centers,  $f(S_i)$

- \* Usual assumption: activity centers equally likely throughout area  $A$
- \*  $f(S_i) = 1/A$
- \* integral usually unwieldy
- \* approximated by numerical approximation using a grid of possible activity centers

• Really nice thing: get estimate of density:  $\hat{N}/A$