

# 1 Materials and Methods

## 1.1 Model specification

$y_{i,j,t}$  is the observed binary occurrence of taxon  $i$  in province  $j$  at time  $t$ , where  $i = 1, 2, \dots, N$ ,  $j = 1, 2, \dots, J$ , and  $t = 1, 2, \dots, T$ .  $y = 1$  is occupied while  $y = 0$  is unoccupied.

$z_{i,j,t}$  is the estimated binary occurrence of taxon  $i$  in province  $j$  at time  $t$ , given the estimate of sampling.  $z = 1$  is occupied while  $z = 0$  is unoccupied.

$p_{j,t}$  is the probability of preservation ( $Pr(y = 1|z = 1)$ ) in province  $j$  at time  $t$ .

$\phi_{j,t}$  is the probability of surviving, in province  $j$ , from time  $t$  to time  $t + 1$  ( $Pr(z_{t+1} = 1|z_t = 1)$ ).

$\gamma_{j,t}$  is the probability of newly entering province  $j$  at time  $t + 1$  ( $Pr(z_{t+1} = 1|z_t = 0)$ ).

$\psi$  is probability of sit occupancy/probability of occurrence ( $Pr(z_{i,t=1} = 1)$ ). The first time point is defined in terms of  $\psi$  because there is (assumed) no previous time points.

$$\begin{aligned} y_{i,t,j} &\sim \text{Bern}(p_{t,j}z_{i,t,j}) \\ z_{i,1,j} &\sim \text{Bern}(\psi_j) \\ z_{i,t,j} &\sim \text{Bern}(\phi_{j,t-1}z_{i,t-1,j} + \gamma_{j,t-1}(1 - z_{i,t-1,j})) \end{aligned} \tag{1}$$

The parameters  $\phi$ ,  $\gamma$ , and  $p$  are then all defined hierarchically within each province, hierarchical by time bin with the mean of that hierarchical by province.

$\Phi_j$ ,  $\Gamma_j$ , and  $P_j$  are the overall probabilities for province  $j$ .  $M_\phi$ ,  $M_\gamma$ , and  $M_p$  are the overall estimates of survival, origination, and preservation probabilities.

$\alpha$  and  $\beta$  are both  $J \times J$  matrices, where  $\beta_j$  would be a column vector.  $X$  is a  $J \times T$  matrix where  $X_t$  is a column vector where each element is defined  $X_{j,t} = \sum_{i=1}^N z_{i,j,t}$  (i.e. the sum of the diversity in province  $j$  at time  $t$ ).

$$\begin{aligned}
\text{logit}^{-1}(\phi_{j,t}) &\sim \text{N}(\Phi_j + \beta_j^t X_{t-1}, \sigma_{\phi,j}) \\
\Phi_j &\sim \text{N}(M_\phi, \sigma_\Phi) \\
\sigma_{\phi,j} &\sim \text{C}^+(1) \\
M_\phi &\sim \text{N}(0, 1) \\
\sigma_\Phi &\sim \text{C}^+(1)
\end{aligned} \tag{2}$$

$$\begin{aligned}
\text{logit}^{-1}(\gamma_{j,t}) &\sim \text{N}(\Gamma_j + \alpha_j^t X_{t-1}, \sigma_{\gamma,j}) \\
\Gamma_j &\sim \text{N}(M_\gamma, \sigma_\Gamma) \\
\sigma_{\gamma,j} &\sim \text{C}^+(1) \\
M_\gamma &\sim \text{N}(0, 1) \\
\sigma_\Gamma &\sim \text{C}^+(1)
\end{aligned} \tag{3}$$

$$\begin{aligned}
\text{logit}^{-1}(p_{j,t}) &\sim \text{N}(P_j, \sigma_{p,j}) \\
P_j &\sim \text{N}(M_p, \sigma_P) \\
\sigma_{p,j} &\sim \text{C}^+(1) \\
M_p &\sim \text{N}(0, 1) \\
\sigma_P &\sim \text{C}^+(1)
\end{aligned} \tag{4}$$

And finally, I use independent priors for  $\psi_j$  by province  $j$ :  $\psi_j \sim \text{U}(0, 1)$ .