

Open population models I: non-spatial models



OPEN POPULATION MODELS

- Open populations experience recruitment or mortality, emigration or immigration.
- N changes across time (seasons, years, or other), $t=1,2,\dots,T$
 $N[t]$ = population size in year t
- We want to estimate $N[t]$ from a model that is jointly specified for all the data. i.e., “multi-year model”

OPEN POPULATION MODELS

■ Two approaches to modeling open systems:

- **Approach 1: Non-dynamic model.** Model the period-specific $N[t]$ parameters as independent parameters. (“implicit dynamics”?)
 - Years are regarded as independent strata.
 - $N[t]$ are independent across years.
 - These approaches ignore individual identity across years and treat recaptures of the same individual as new individuals.
 - The approach is therefore less efficient for estimating $N[t]$ but often not by much.
- **Approach 2: Dynamic model.** (“explicit dynamics”) Preserves individual identity – more efficient use of data. Model transitions of individuals from one year to the next.
 - $N[t]$ are not independent across years.
 - Allows estimation of dynamics parameters.
 - JS and CJS type models

NON-DYNAMIC OPEN MODELS

Two approaches to implementation:

- **Year as an individual class variable:**

 - `year[i] ~ dcat(probs[year])`

 - This allocates the uncaptured individuals into years
 - Regards individuals as independent across years (recaptures are new individuals)
 - We already did this on Day 1

- **T-fold data augmentation.** Analyze T data sets, each one augmented individually with DA parameter $\psi[t]$.
Operationally: Input the data as a 3-d array "individual x replicate x year"

 - Also regards individuals as independent across years

- These are practically equivalent if $\dim(\text{probs}[\text{year}]) = T-1$

OPEN POPULATIONS

- Both classes of models are easily 'spatialized'. i.e., once we have the ordinary non-spatial model described in BUGS, then we can easily add the latent activity centers $s[i]$ with a few lines of code
- Therefore we will spend a fair amount of time talking about the **non-spatial** versions of these models.

HOW TO DO THIS IN BUGS

T-fold data augmentation:

Create T data sets, augmented to the same dimension M (for convenience)

- **3-d array data structure:** Store the data in an $M \times K \times T$ array with elements $y[i,k,t] = 1$ if captured during occasion k year t, etc..
- **2-d version based on summary statistics.** If there are no effects that vary in the k-dimension (behavioral response, occasion-specific parameters) then we can summarize the data to $y[i,t] = \text{total number of captures out of } K$

T-FOLD DATA AUGMENTATION WITH THE 3-D ARRAY

```
for(t in 1:5){
  N[t] <- sum(z[ ,t])
  psi[t] ~ dunif(0,1)
  p[t] ~ dunif(0,1)
}
for(t in 1:5){
  for(i in 1:M){
    z[i,t] ~ dbern(psi[t])
    for(k in 1:K){
      mu[i,k,t] <- z[i,t]*p[t]*(1-isdead[i,k,t])
      Yarr[i,k,t] ~ dbern( mu[i,k,t] )
    }
  }
}
```

T-FOLD DATA AUGMENTATION: DATA SUMMARIZED TO 2-D ARRAY

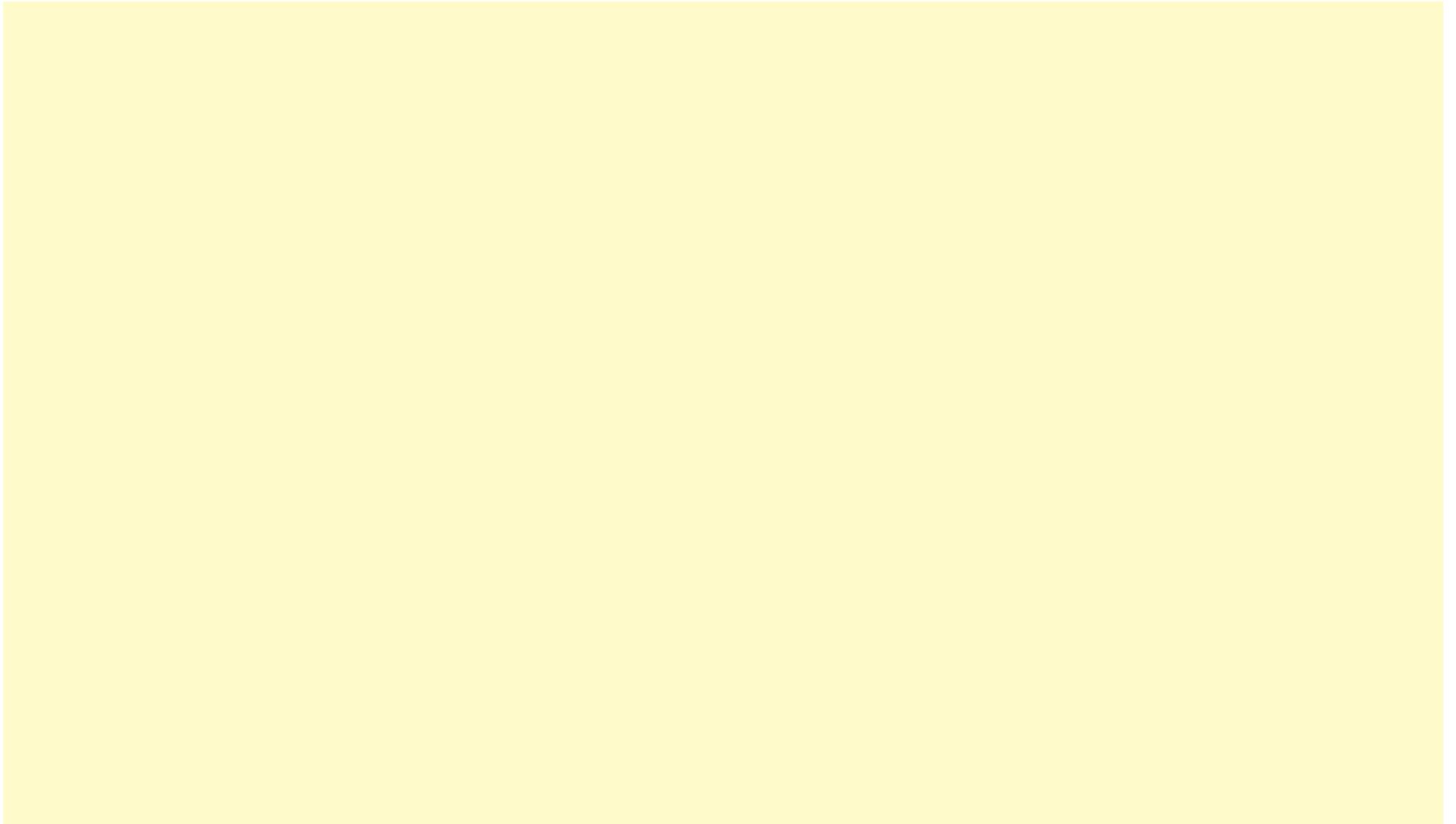
```
model {  
  
  for(t in 1:5){  
    N[t] <- sum(z[ ,t])  
    psi[t] ~ dunif(0,1)  
    p[t] ~ dunif(0,1)  
  }  
  
  for(t in 1:5){  
    for(i in 1:M){  
      z[i,t] ~ dbern(psi[t])  
      mu[i,t] <- z[i,t]*p[t]  
      Y2d[i,t] ~ dbin( mu[i,t], Kmat[i,t] ) # Kmat[i,t] instead of isdead  
    }  
  }  
}
```

T-FOLD DATA AUGMENTATION

- R work session

TOWARD A DYNAMIC MODELS

- These approaches do not preserve individual ID across years
- It is possible to extend the “year as a class variable” specification to include individual ID but it is inelegant and untidy.
- The 3-d array or “stacked” T-fold DA models can be extended to preserve individual ID
- Ignoring individual identity across years, treating recaptures as new individuals, disregards a small bit of information and thus we’re making inefficient use of data, not getting the most efficient estimates (in a variance sense), and also not able to estimate survival and recruitment



PRESERVING INDIVIDUAL IDENTITY IN OPEN MODELS

Population model with survival and recruitment is **exactly a restricted occupancy model** (Ch. 10 Royle and Dorazio 2008). Recruitment = colonization, 1-extinction = survival. The restriction is: once an animal dies it can't be recruited again.

If an individual is currently alive:

$$z[i,t] \sim \text{Bern}(\text{phi} * z[i,t-1])$$

If an individual is currently not yet recruited prior to t

$$z[i,t] \sim \text{Bern}(\text{gamma} * z[i,t-1])$$

To implement this model we need to keep track of whether an animal was ever alive, so that it cannot be re-recruited....

We take code directly from Kery and Schaub (2012, Ch. 10).

BUGS CODE FOR JOLLY-SEBER MODEL

```
# Priors and constraints
for (i in 1:M){
  for (t in 1:( T-1 )){
    phi[i,t] <- mean.phi
  } #t
  for (t in 1:T){
    p[i,t] <- mean.p
  } #t
} #i
mean.phi ~ dunif(0, 1)
mean.p ~ dunif(0, 1)
for (t in 1:T){
  gamma[t] ~ dunif(0, 1)      # year-specific colonization probability
  N[t] <- sum(z[1:M,t])      # Actual population size
} #t

# Observation model
for (i in 1:M){
  for(t in 1:T){
    mu1[i,t] <- z[i,t] * p[i,t]
    y[i,t] ~ dbin(mu1[i,t],Kmat[i,t])
  }
}

## Process model
for(i in 1:M){
  # Initial state
  z[i,1] ~ dbern(gamma[1])
  # Subsequent occasions
  for (t in 2:T){
    q[i,t-1] <- 1-z[i,t-1]          # Availability for recruitment
    available[i,t-1]<- prod(q[i,1:(t-1)])
    mu2[i,t] <- phi[i,t-1] * z[i,t-1] + gamma[t] * available[i,t-1]
    z[i,t] ~ dbern(mu2[i,t])
  } #t
} #i
```

Constant phi and p here but could vary by i and t easily

JS MODEL: RESTRICTED OCCUPANCY MODEL

- Useful derived parameters:
 - $N[t]$ = number of occupied sites in period t
 - $B[t]$ = number of newly occupied sites in period t

Open population models II: SCR models



OPEN SCR MODELS

Two types of “open” dynamics we can model:

1. **Population dynamics**: individuals are entering and leaving the population through recruitment and mortality
2. **Spatial dynamics** – dynamic activity centers (dispersal, migration, transience)

Open **SCR** models can have one or both things happening

CLASSES OF “OPEN” SCR MODELS

- **Closed SCR model** with dynamic activity centers
- Implicit dynamics (“N random”), static activity centers
- Explicit dynamics model (Jolly-Seber), static activity centers
- Explicit dynamics model with dynamic activity centers

“open population models” should be “dynamic population models” and we have different types of dynamics acting on different components of the model.

5 (OR 6) CLASSES OF “OPEN” SCR MODELS

SPATIAL DYNAMICS

POPULATION DYNAMICS

| | s[i] static | s[i] dynamic |
|------------------|---|-------------------------------|
| None | Basic SCR model (not open) | SCR + dispersal or transience |
| Random N[t] | “year as a class variable” s[i,t] ~ random | [no individual ID] |
| Partial dynamics | Cormack-Jolly-Seber | + s[i,t] |
| Fully dynamic | Jolly-Seber model | + s[i,t] |

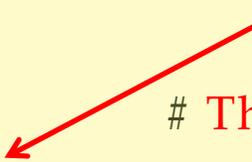
“OPEN” SCR I: CLOSED SCR MODEL WITH SPATIAL DYNAMICS

- Ordinary SCR model with transience or dispersal
- Accounts for non-closure due to animals moving away from the area of the state-space where they were catchable.
- Accounts for correlated space usage by individuals
- Modifications to the model:
 - $s[i,k]$ instead of $s[i]$
 - Model for $s[i,k]$

Plausible models:

1. $s[i,k] \sim \text{Uniform}(S)$ # This model is not estimable
2. $s[i,k] \sim \text{normal}(\mu[i], \text{sigma}^2)$ # $\mu[i]$: meta-homerange center
3. $s[i,k] \sim \text{normal}(s[i,k-1], \text{sigma}^2)$ # Markovian transience

Different “sigma” here



CLOSED SCR MODEL WITH SPATIAL DYNAMICS

- For the Markovian dispersal or transience models we need to increase the size of the state-space to allow sufficient area for movement (relative to size of movement scale parameter)
- Also have to preserve density of the state-space – individuals can't leave. It is easiest to truncate the movement/dispersal model to restrict $s[i,k]$
- **Demonstration using the Fort Drum black bear data**

OPEN SCR II: IMPLICIT DYNAMICS, RANDOM ACTIVITY CENTERS

- Here we spatialize that model in two ways:
 - 1. Including static activity centers $s[i]$ – suggesting that the home ranges are stationary over time
 - 2. by including activity centers for each individual and year, $s[i,t]$
 - “secr” can fit model 2 using the multi-session approaches.
- These models accommodate implicit dynamics and possible changes in activity centers

OPEN SCR III: SPATIAL JOLLY-SEBER TYPE MODEL

- Variations of this model:
 1. JS model with static activity centers
 2. JS model with random activity centers
 3. JS model with dynamic activity centers

SUMMARY

- Many variations of “non-closure” in SCR models:
 - Spatially dynamic activity centers
 - Population dynamics (survival and recruitment)
- Can have spatial dynamics operating even in closed populations
- A benefit of using BUGS is all of these models are relatively easy to parameterize and analyse.
- In ‘secur’ you can only do limited things.