

### Lecture 15

Multi-state models

WILD6900 (Spring 2020)

# Readings

Kéry & Schaub 264-313

### What is a "state"

#### Alive/dead

CJS model

### Occupied/unoccupied

Occupancy model

Geographic location (location  $1, 2, \ldots, S$ )

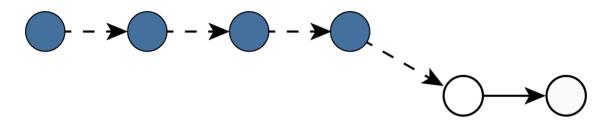
Un-infected/infected/dead

Age class

Any combination of above states\*

### Two states

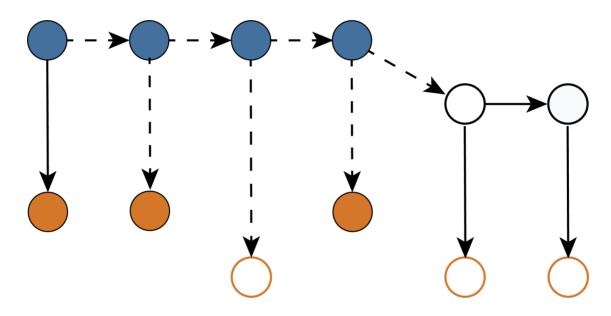
State model



$$z_{i,t} \sim Bernoulli(z_{i,t-1}\phi)$$

### Two states

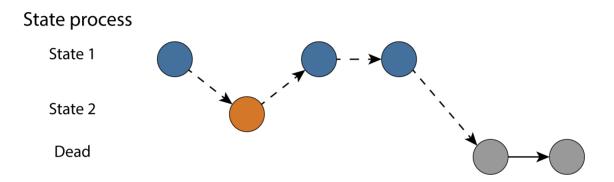
### Observation model



$$y_{i,t} \sim Bernoulli(z_{i,t}p)$$

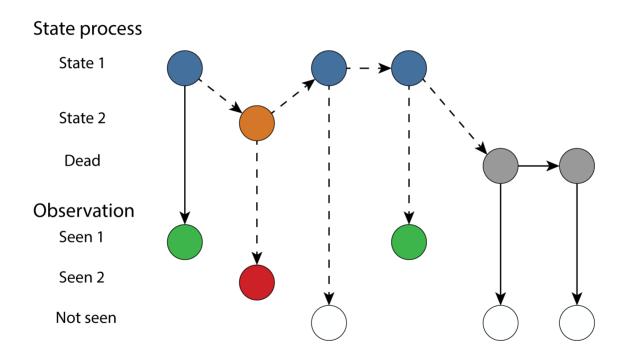
## Three states

### State model



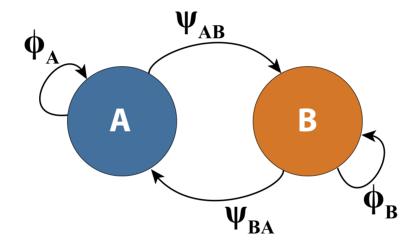
## Three states

### State model



Movement between two sites

Movement between two sites



- $\phi_A$ : Probability alive at site A
- $\phi_B$ : Probability alive at site B
- ullet  $\psi_{AB}$ : Probability of moving from site A to site B
- ullet  $\psi_{BA}$ : Probability of moving from site B to site A

State model

#### True state at time t + 1

True state at time *t* 

	Site A	Site B	Dead
Site A	$\phi_A(1-\psi_{AB})$	$\phi_A\psi_{AB}$	$1-\phi_A$
Site B	$\phi_B\psi_{BA}$	$\phi_B(1-\psi_{BA})$	$1-\phi_B$
Dead	0	0	1

### Observation model

#### Observation at time t

True state at time *t* 

	Site A	Site B	Not Seen
Site A	$p_A$	0	$1-p_A$
Site B	0	$p_B$	$1-p_B$
Dead	0	0	1

#### The data

- 1 = Seen alive at site A
- 2 = Seen alive at site B
- 3 = Not seen

#### head(rCH)

```
## [,1] [,2] [,3] [,4] [,5] [,6]

## [1,] 1 1 3 1 1 1 1

## [2,] 1 1 1 1 1 1 1 1

## [3,] 1 3 1 2 3 3

## [4,] 1 3 3 3 3 3 3

## [5,] 1 1 2 3 3 3

## [6,] 1 3 3 2 3 1
```

#### The data

• f =First capture occasion

```
tail(rCH)
##
  [,1] [,2] [,3] [,4] [,5] [,6]
## [795,] 3 3
## [796,] 3 3 3 2 3
## [797,] 3 3 3 3 2 3
## [798,] 3 3 3 2 3
## [799,] 3 3 3 2 1
## [800,]
tail(f)
## [1] 5 5 5 5 5 5
```

#### The likelihood

- The Bernoulli distribution describes the probability of one (of two) outcomes
- ullet The categorical distribution describes the probability of one (of N) outcomes
  - $\circ \Omega$  and  $\Theta$  are matrices defining the state/observation probabilities

$$|z_{i,t+1}|z_{i,t} \sim categorical(\Omega_{z_{i,t},1:S})$$

$$y_{i,t}|z_{i,t} \sim categorical(\Theta_{z_{i,t},1:S})$$

```
sink("ms.jags")
cat("
model {
    # Priors
    phiA ~ dunif(0, 1)
    phiB ~ dunif(0, 1)
    psiAB ~ dunif(0, 1)
    psiBA ~ dunif(0, 1)
    pA ~ dunif(0, 1)
    pB ~ dunif(0, 1)
```

```
# Define state-transition and observation matrices
# Define probabilities of state S(t+1) given S(t)

ps[1,1] <- phiA * (1-psiAB)

ps[1,2] <- phiA * psiAB

ps[1,3] <- 1-phiA

ps[2,1] <- phiB * psiBA

ps[2,2] <- phiB * (1-psiBA)

ps[2,3] <- 1-phiB

ps[3,1] <- 0

ps[3,2] <- 0

ps[3,3] <- 1</pre>
```

```
# Define probabilities of O(t) given S(t)
po[1,1] <- pA
po[1,2] <- 0
po[1,3] <- 1-pA
po[2,1] <- 0
po[2,2] <- pB
po[2,3] <- 1-pB
po[3,1] <- 0
po[3,2] <- 0
po[3,3] <- 1</pre>
```

```
# Likelihood
for (i in 1:nInd){
   # Define latent state at first capture
   z[i,f[i]] <- y[i,f[i]]
   for (t in (f[i]+1):n0cc){
      # State process: draw S(t) given S(t-1)
      z[i,t] \sim dcat(ps[z[i,t-1],])
      # Observation process: draw O(t) given S(t)
      y[i,t] \sim dcat(po[z[i,t],])
      } #t
   } #i
",fill = TRUE)
sink()
```

### Multi-state models

### Very flexible

- Most capture-recapture models can be formulated as multi-state models
- Can accommodate many states
- Can include covariates and individual/temporal variation

### But complexity comes at a cost

- Data hungry
- Often not fully identifiable
- Computationally intensive