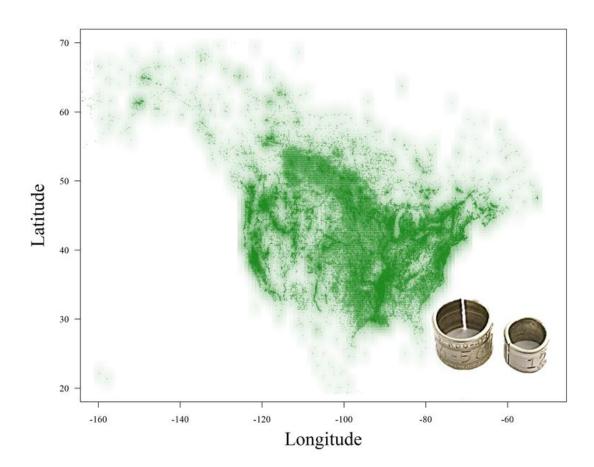
Bayesian band-recovery model workshop

Thomas Riecke, *University of Montana*Madeleine Lohman, *University of Nevada*Benjamin Sedinger, *University of Wisconsin-Stevens Point*Jordan Thompson, *Colorado State University*Caroline Blommel, *Colorado State University*Dan Gibson, *University of Minnesota*



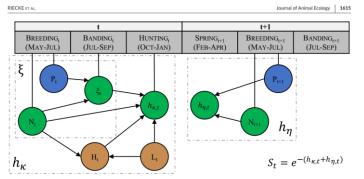


FIGURE 2 Directed acyclic graph demonstrating the hypothesized relationships among mallard breeding pair advandance (N), the number of ponds (P), harvest limits (L), the abundance of ulculo hunters (H), fecultive (V), harvest mortality hazard rate (P), and survival (S) for mallards marked and released in the Prairie Pothole Region of the United States and Canada, 1974-2016. Arrows represent covariate effects, grey dashed lines enclose separate generalized linear models and vertical solid lines denote the time period or interval when parameters were estimated, where survival (S) and natural mortality in year at are estimated from banding in year to 1 as dimained from banding in year to 1 as dimaining in year to 1 as dimained from banding in year to 1 as dimaining in year to 1 as dimained promoting in year to 1 as dimained promoting in year to 1 as dimaining in year to 1 as dimained promoting in year to 1 as dimaining in year to 1 as dimainin

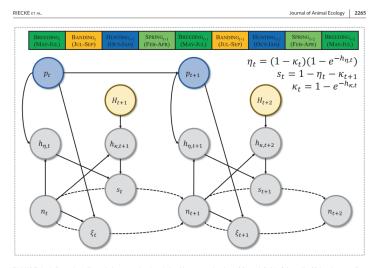


FIGURE 2 A directed acyclic graph demonstrating the relationships among abundance (n), ponds (n; blue), fecundity (2), hunting mortality hazard rate (n), hatural mortality hazard rate (n), survival (s) and the number of duck hunters (H; brown) for blue-winged teal breeding in the North American Pairie Pothole Region across the annual cycle (1973–3016). Solid arrows represent estimated directional relationships, and dashed arrows represent processes leading to changes in population abundance.

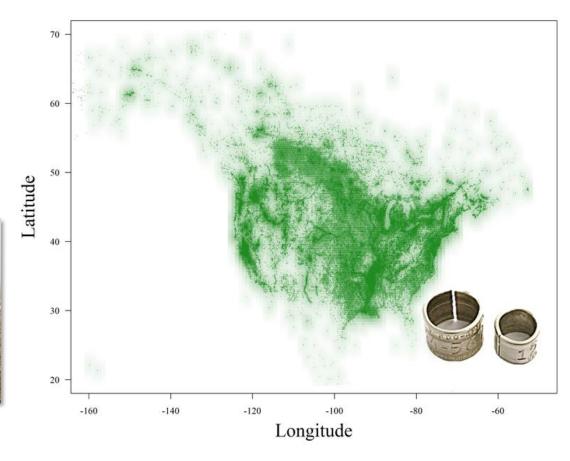
These data are the product of citizen-science











Cavell Brownie figured this out 50 years ago!





Dr. Cavell Brownie
PhD, Cornell University
Professor Emeritus, North Carolina State University



Going Bayesian







Todd Arnold



Scott Boomer



Jim Sedinger

Facilitators



Thomas Riecke & Ben Sedinger



Dan Gibson



Madeleine Lohman



Caroline Blommel



Jordan Thompson

Workshop 'tone'

- 1. The facilitators are the strength of the workshop.
- 2. We're going to cover concepts at different levels: ask questions!
- 3. We're going to do small group exercises: get to know folks!
- 4. This is complicated: be patient (and kind).

Outline

- 1. How band-recovery models work and some coding basics (1h)
- 2. Priors: estimating band-recovery and harvest probability (1h)
- 3. Survival models: random effects and covariates (1h)
- 4. Lincoln estimates and cross-seasonal models (1h)
- 5. Path analysis and acquiring data from the BBL (1h)

1. How band-recovery models work and coding basics

Year banded	Number - banded	Recoveries by hunting season				
		1	2	3,		,k
1	N_1	R_{11}	R_{12}	R_{13}		R_{1i}
2	N_z		R_{zz}	R_{23} .	***	R_{24}
3	N_3		1000	R_{33} ,	***	R_{ik}
					*	
					*	
					*	
k	N_k					$R_{i,i}$

How do we format our data...?!

An example from a 2-year study

$$M = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

$$R = \begin{bmatrix} \mathbf{1000} \\ 1000 \end{bmatrix}^{2}$$

We released 1000 birds in year 1

*m-array is short for multinomial array

$$M = \begin{bmatrix} 1000 \\ \mathbf{1000} \end{bmatrix}$$

$$R = \begin{bmatrix} 1000 \\ \mathbf{1000} \end{bmatrix}$$

We released 1000 birds in year 2

$$M = \begin{bmatrix} \mathbf{1000} \\ \mathbf{1000} \end{bmatrix}$$

$$R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

We had 100 direct recoveries in year 1

$$M = \begin{bmatrix} 100 & \mathbf{50} \\ 1000 \end{bmatrix}$$

$$R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

We had 50 <u>indirect</u> recoveries initially marked in year 1 recovered in year 2

$$M = \begin{bmatrix} 100 & 50 & 850 \\ 1000 \end{bmatrix} \qquad R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

850 birds from year 1 were never recovered

$$M = \begin{bmatrix} 100 & 50 & 850 \\ 1000 \end{bmatrix} \qquad R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

'Fate(s)' of 1000 birds released in year 1...

$$M = \begin{bmatrix} 100 & 50 & 850 \\ \mathbf{0} & & & \end{bmatrix} \qquad R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

No birds released in year 2 were shot in year 1

$$M = \begin{bmatrix} 100 & 50 & 850 \\ 0 & \mathbf{100} \end{bmatrix} \qquad R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

We had 100 direct recoveries in year 2

$$M = \begin{bmatrix} 100 & 50 & 850 \\ 0 & 100 & 900 \end{bmatrix} \qquad R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

900 birds from year 2 were never recovered

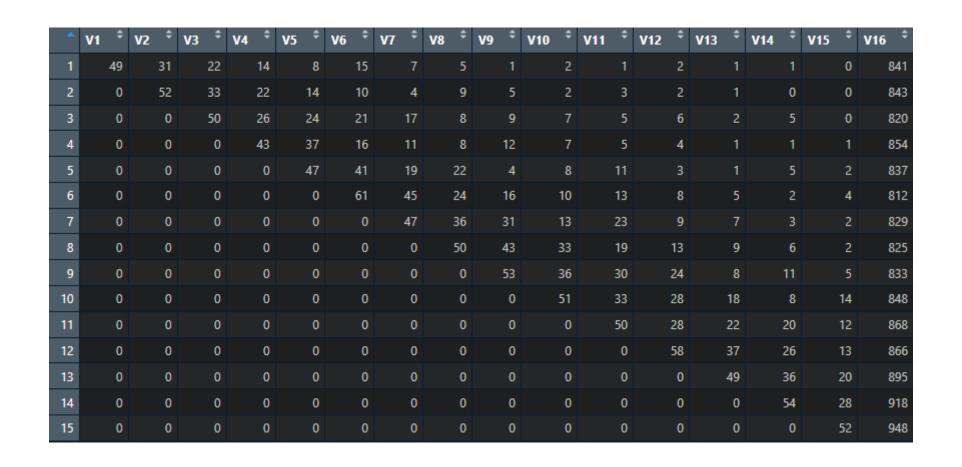
$$M = \begin{bmatrix} 100 & 50 & 850 \\ 0 & 100 & 900 \end{bmatrix} \qquad R = \begin{bmatrix} 1000 \\ \mathbf{1000} \end{bmatrix}$$

'Fate(s)' of 1000 birds released in year 2...

$$M = \begin{bmatrix} \mathbf{100} & \mathbf{50} & \mathbf{850} \\ 0 & 100 & 900 \end{bmatrix} \qquad R = \begin{bmatrix} \mathbf{1000} \\ 1000 \end{bmatrix}$$

Each row of M has to add up to that year's R

This starts to get complicated (?!)



Oh geez... we'll learn some simple tricks!

Our first (back-of-the-envelope) survival estimate!

An example from a 2-year study

The basic idea: releases (R)

 Let's imagine we capture, mark, and release 1000 ad. male mallards (immediately prior to the hunting season)

$$R_1 = 1000$$



$$R = \begin{bmatrix} 1000 \\ \end{bmatrix}$$

$$M = \left[\begin{array}{c} \\ \end{array} \right]$$

The basic idea: m-arrays (M: multinomial array)

- Let's imagine we capture and mark 1000 ad. male mallards
- 100 are shot and reported by hunters that hunting season
 - i.e., direct 'recoveries'



$$R = \begin{bmatrix} 1000 \\ \end{bmatrix}$$

$$m_{1,1} = 100$$

$$M = \begin{bmatrix} \mathbf{100} \\ \end{bmatrix}$$

The basic idea: band recovery probability (f)

- Let's imagine we capture and mark 1000 ad. male mallards
- 100 are shot and reported by hunters that hunting season
 - i.e., 'direct' recoveries

$$f_1 = \frac{100}{1000} = 0.1$$



P: cell probabilities

$$P = \int_{1}^{f_1}$$

$$R = \begin{bmatrix} 1000 \\ \end{bmatrix}$$

$$M = \begin{bmatrix} 100 \\ \end{bmatrix}$$

The basic idea

 The exact same thing happens next year (seems unlikely, but ok...)

$$P = \begin{bmatrix} f_1 \\ 0 & f_2 \end{bmatrix}$$

$$R = \begin{bmatrix} 1000 \\ \mathbf{1000} \end{bmatrix}$$

$$f_2 = \frac{100}{1000} = 0.1$$



$$M = \begin{bmatrix} 100 \\ 0 & \mathbf{100} & 900 \end{bmatrix}$$

A note about subscripts

$$f_1 = \frac{100}{1000} = 0.1$$
 This corresponds to year 1

$$f_2 = \frac{100}{1000} = 0.1$$

This corresponds to year 2

Imagine we also shoot some more birds marked in *t*=1 during the t=2 hunting season

- We also shoot 50 ducks marked in the first year in year 2...
 - i.e., 'indirect' recoveries
- What does the 50 tell us?*

$$P = \begin{bmatrix} f_1 & ? \\ 0 & f_2 \end{bmatrix}$$

$$R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

$$M = \begin{bmatrix} 100 & \mathbf{50} & 850 \\ 0 & 100 & 900 \end{bmatrix}$$

*This is the key idea

If we know we're shooting 10% of the birds....

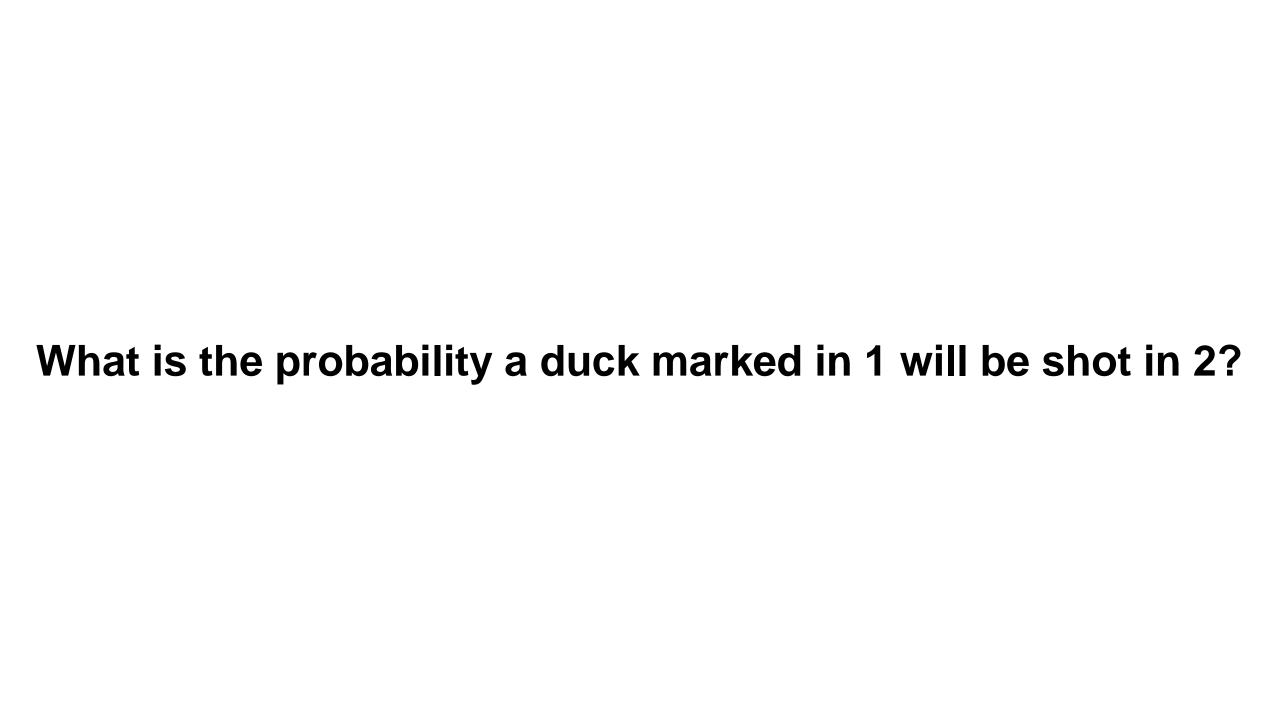
$$100 = 1000 \times f_2$$

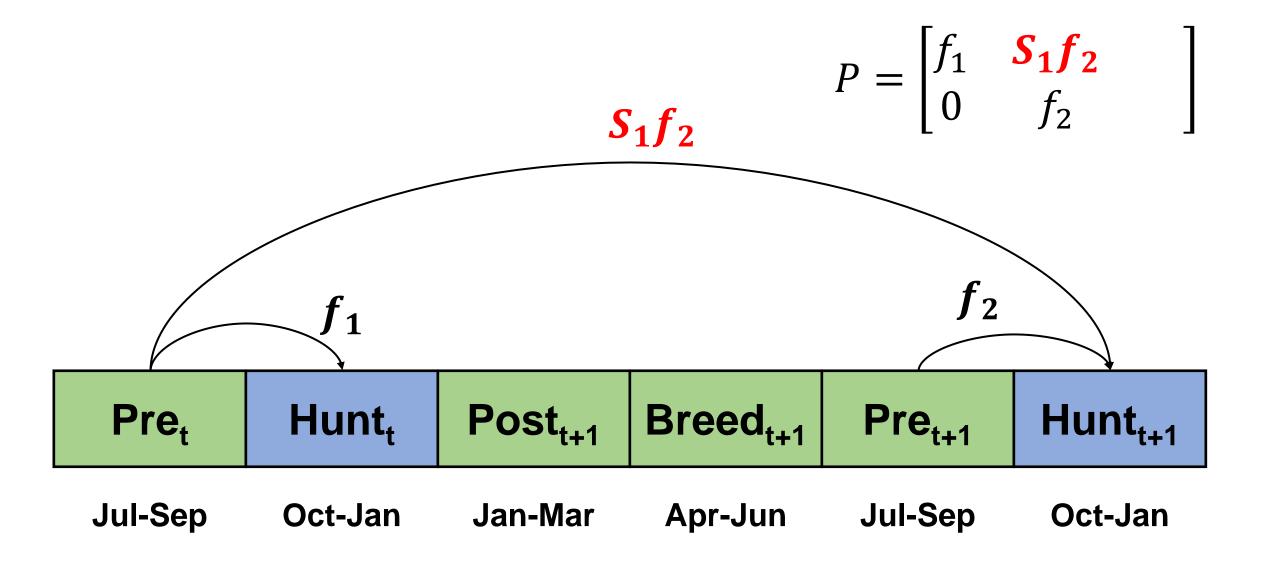
 $f_2 = 0.1$

$$P = \begin{bmatrix} f_1 \\ 0 & f_2 \end{bmatrix}$$

$$R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

$$M = \begin{bmatrix} 100 & \mathbf{50} & 850 \\ 0 & \mathbf{100} & 900 \end{bmatrix}$$





What is the probability a duck marked in 1 will be shot in 2?

- It has to survive one year (S₁)
- And be recovered in the next (f₂)

$$P = \begin{bmatrix} f_1 & \mathbf{S_1 f_2} \\ 0 & f_2 \end{bmatrix}$$

What does the 50 tell us?

$$R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

$$M = \begin{bmatrix} 100 & \mathbf{50} & 850 \\ 0 & 100 & 900 \end{bmatrix}$$

Imagine survival was 50%...

We released 1000 birds a year ago...

How many are still alive?

Imagine survival was 50%...

We released 1000 birds a year ago...

How many are still alive? 500

Now imagine we shoot 10% of those 500 birds...

How many would we shoot?

Now imagine we shoot 10% of those 500 birds...

How many would we shoot? 50

We estimate survival as a function of R and f

$$100 = R_2 \times f_2$$

 $100 = 1000 \times f_2$
 $f_2 = 0.1$

$$50 = R_1 \times S_1 \times f_2$$

$$50 = 1000 * S_1 * 0.1$$

$$S_1 = \frac{50}{1000 \times 0.1} = 0.5$$

$$P = \begin{bmatrix} f_1 & \mathbf{S_1} f_2 \\ 0 & f_2 \end{bmatrix}$$

$$R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

$$M = \begin{bmatrix} 100 & \mathbf{50} & 850 \\ 0 & \mathbf{100} & 900 \end{bmatrix}$$

50 is 10% of <u>500</u>

If we shoot 50 birds banded in the first year in year two, and we know we're shooting 10% of the ducks, then how many were available to be shot?

If 500 are alive (available to be shot) now, and 1000 were alive one year ago, what is S?

$$S_1 = \frac{50}{1000 \times 0.1} = 0.5$$
 $f_2 = \frac{100}{1000} = 0.1$

What would our survival estimate be if...

we'd had 60 indirect recoveries?

we'd had 70 indirect recoveries?

we'd had 70 recoveries, but $f_2 = 0.2$?

That's the basic idea.

We know what proportion we shoot.

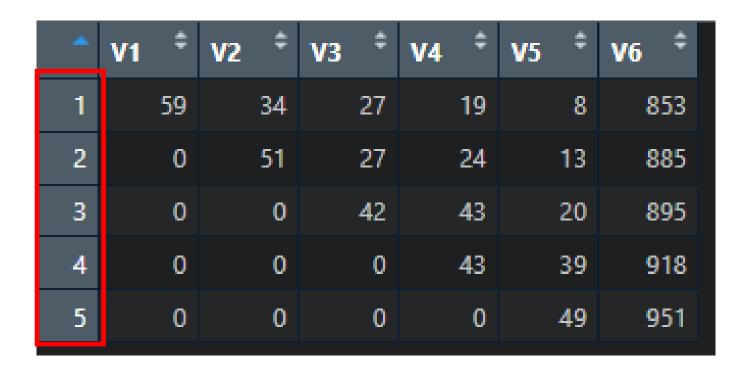
The number of indirect recoveries tell us number still alive

If we know the number still alive, we can estimate S!

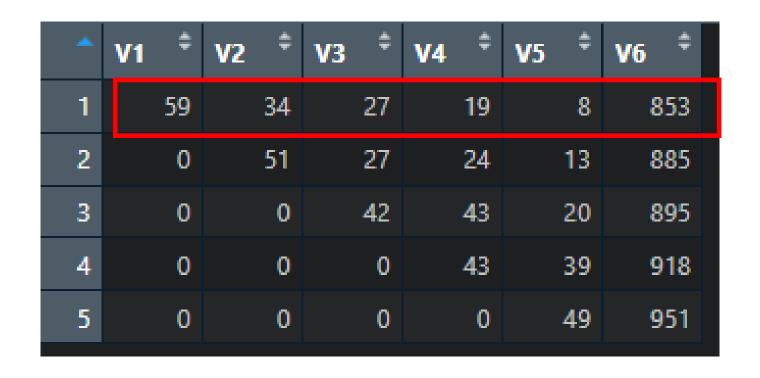
This starts to get complicated (?!)

*	V1 [‡]	V2 [‡]	V 3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

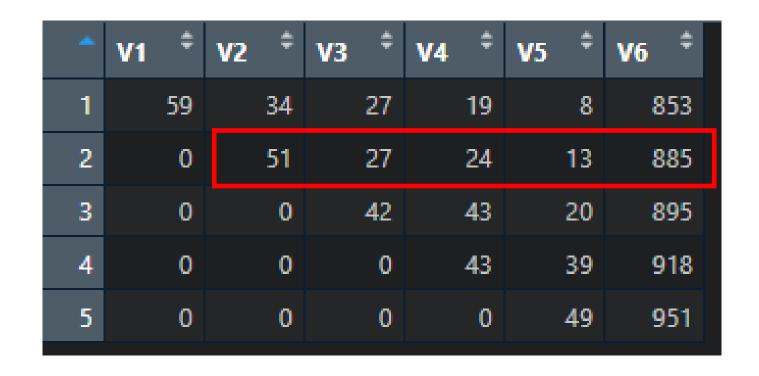
Oh geez...



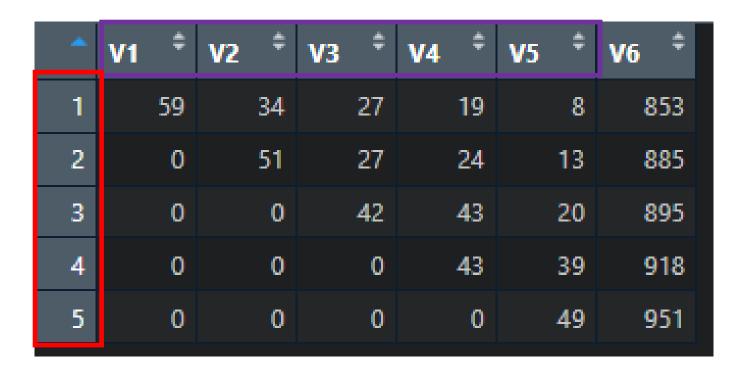
The rows correspond to year of release



This is what happened to birds released in the first year



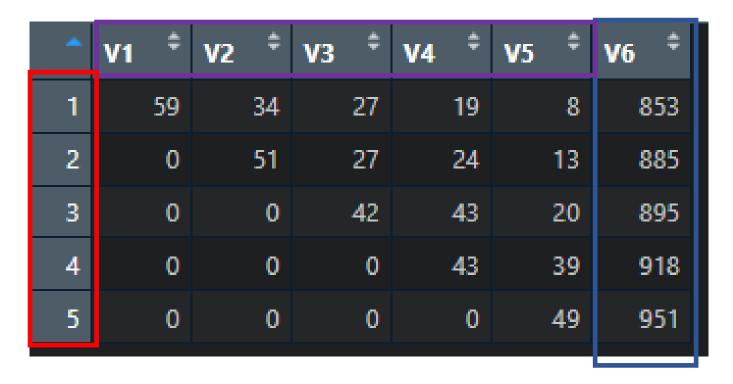
This is what happened to birds released in the second year



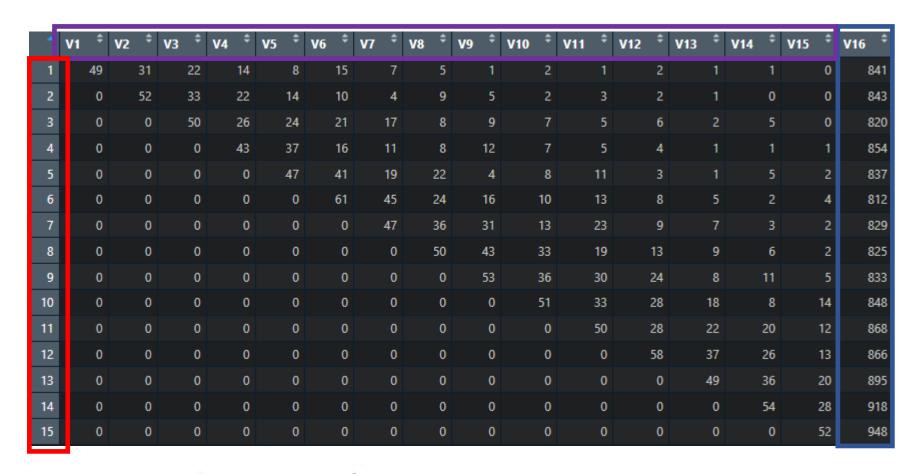
The rows correspond to year of release
The columns correspond to year of recovery

	V1 [‡]	V2 [‡]	V3 [‡]	V4 [‡]	V5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

These are recoveries from year 4



Year of recovery
$$M = \begin{bmatrix} 100 & 50 & 850 \\ 0 & 100 & 900 \end{bmatrix}$$



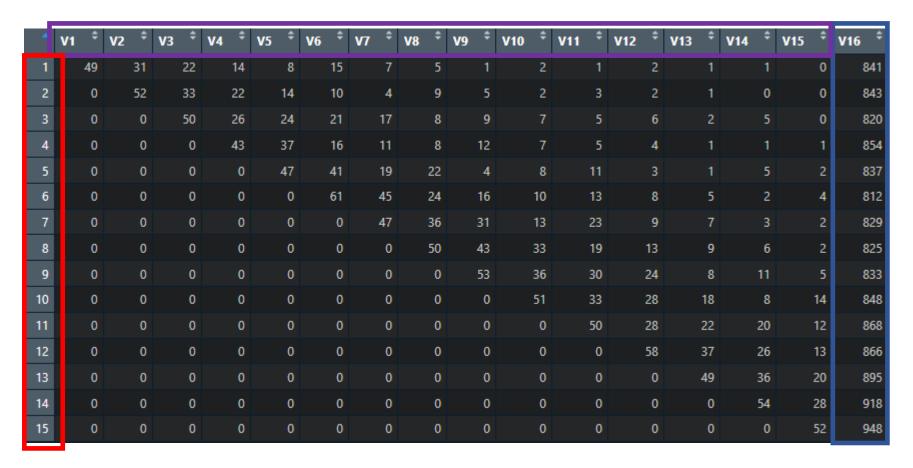
It's fairly simple to do this by hand with tiny datasets

$$P = \begin{bmatrix} f_1 & S_1 f_2 & 1 - f_1 - S_1 f_2 \\ 0 & f_2 & 1 - f_2 \end{bmatrix} \qquad R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

$$R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

$$M = \begin{bmatrix} 100 & 50 & 850 \\ 0 & 100 & 900 \end{bmatrix}$$

How do we deal with this amount of info (240 cells)?



Understanding indexing is key to writing code...





Refuge[big shop, east wall, 3rd shelf from right, behind the nets, little case]

1-dimensional indexing!

$$R_1 = 1000$$

$$R = \begin{bmatrix} 1000 \\ 950 \end{bmatrix}$$

data[#]

2-dimensional indexing!

As simple as a spreadsheet

*	V1 [‡]	V2 [‡]	V3 [‡]	V4 [‡]	V 5 [‡]	V 6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

data[row, column]

M-array[year released, year recovered]

3-dimensional indexing! As simple as a notebook

m[1, 4, 1]; m_{1,4,1}

*	V1 [‡]	V2 [‡]	V 3 [‡]	V4 ‡	V5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

m[3, 3, 2]; m_{3,3,2}

•	V1 [‡]	V2 ‡	V 3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

data[row, column, page]

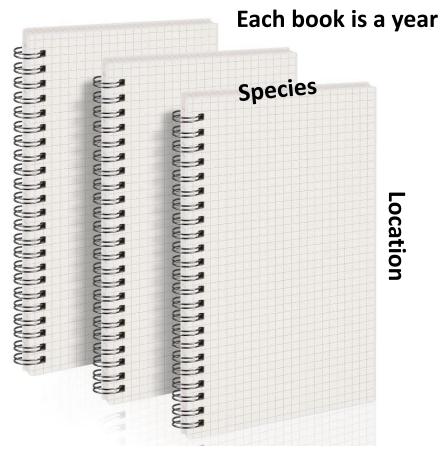
M-array[year released, year recovered, sex]



4-dimensional indexing! (a stack of notebooks) imagine surveys at twenty locations (L) for fifteen species (S) that are performed five times per breeding season (J) across six years (T)

```
row: location [1,2,3,...,L] col: species [1,2,3,...,S] page: survey [1,2,3,...,J] book: year [1,2,3,...,T]
```

data[row, column, page, data book] data[location, species, survey, year]



Turn the page for next survey

7-dimensional indexing! (walmart) Back-to-school sale





Walmart[row, column, page, book, shelf row, shelf column, aisle]
Walmart[1, 1, 1, 1, 1, end cap aisle 24/25]

19 ducks were released in year 1 and shot in year 4

V1-V5: Year of recovery V6: never reencountered

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•	V1 [‡]	V2 [‡]	V3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

$$M[1,4] = 19$$

49 ducks were released in year 5 and shot in year 5

V1-V5: Year of recovery V6: never reencountered

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•	V1 [‡]	V2 [‡]	V3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

$$M[5,5] = 49$$

8 ducks were released in year 1 and shot in year 5

V1-V5: Year of recovery V6: never reencountered

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*	V1 [‡]	V2 [‡]	V3 [‡]	V4 [‡]	V5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

$$M[1,5] = 8$$

895 ducks were released in year 3 and never recovered

V1-V5: Year of recovery V6: never reencountered

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•	V1 [‡]	V2 [‡]	V3 [‡]	V4 [‡]	V 5 [‡]	V 6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

$$M[3,6] = 895$$

Why are these all 0's

V1-V5: Year of recovery

V6: never reencountered

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*	V1 [‡]	V2 [‡]	V3 [‡]	V4 [‡]	V 5 [‡]	V 6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

M[2:5,1] = [0,0,0,0]

In R, 2:5 is the same as 2,3,4,5

Why are these all 0's?

V1-V5: Year of recovery

V6: never reencountered

^	V1 ‡	V2 [‡]	V3 [‡]	V4 [‡]	V5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

You can't be shot in year 1 if you were released after t=1...

Why are these the biggest numbers (typically)?

V1-V5: Year of recovery V6: never reencountered

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*	V1 [‡]	V2 ‡	V 3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

M[1:5,6] = [853,885,895,918,951]

Why are these the biggest numbers (typically)?

V1-V5: Year of recovery V6: never reencountered

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*	V1 [‡]	V2 [‡]	V3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

Most ducks aren't shot, retrieved, and reported by a hunter

Why do these numbers (generally) decline?

V1-V5: Year of recovery

V6: never reencountered

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Year
1-5:

*	V1	‡	V2 ‡	V 3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1		59	34	27	19	8	853
2		0	51	27	24	13	885
3		0	0	42	43	20	895
4		0	0	0	43	39	918
5		0	0	0	0	49	951

M[1,1:5] = [59,34,27,19,8]

Why are these the 2nd biggest numbers (typically)?

V1-V5: Year of recovery V6: never reencountered

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^	V1 [‡]	V2 [‡]	V 3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

$$M[1,1] = 39, M[2,2] = 51, M[3,3] = 42$$

Why are these the 2nd biggest numbers (typically)?

V1-V5: Year of recovery V6: never reencountered

V4 **V3** V1 Year of release -5:

These are recoveries immediately after release (no mortality)

What are the probabilities of these outcomes?

V1-V5: Year of recovery V6: never reencountered

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*	V1 [‡]	V2 [‡]	V 3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

What is the probability of this outcome?

V1-V5: Year of recovery

V6: never reencountered

Ф	^	V1 [‡]	V2 [‡]	V 3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
release	1	59	34	27	19	8	853
	2	0	51	27	24	13	885
Year of	3	0	0	42	43	20	895
Yea	4	0	0	0	43	39	918
1-5:	5	0	0	0	0	49	951

Released in year 1, shot in year 2...

What is the probability of this outcome?

V1-V5: Year of recovery

V6: never reencountered

1-5: Year of release

_	V1 [‡]	V2 [‡]	V 3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

You have to survive year 1, and be shot in year 2 (S₁f₂)

What is the probability of this outcome?

$$P = \begin{bmatrix} f_1 & S_1 f_2 & S_1 S_2 f_3 & S_1 S_2 S_3 f_4 & S_1 S_2 S_3 S_4 f_5 & 1 - \sum_{j=1}^{n} P_{1,1:5} \\ 0 & f_2 & S_2 f_3 & S_2 S_3 f_4 & S_2 S_3 S_4 f_5 & 1 - \sum_{j=1}^{n} P_{2,1:5} \\ 0 & 0 & f_3 & S_3 f_4 & S_3 S_4 f_5 & 1 - \sum_{j=1}^{n} P_{3,1:5} \\ 0 & 0 & 0 & f_4 & S_4 f_5 & 1 - \sum_{j=1}^{n} P_{4,1:5} \\ 0 & 0 & 0 & 0 & f_5 & 1 - \sum_{j=1}^{n} P_{5,1:5} \end{bmatrix}$$

V1-V5: Year of recovery

V6: never reencountered

Ф	*	V1 [‡]	V2 ‡	V3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
release	1	59	34	27	19	8	853
	2	0	51	27	24	13	885
Year of	3	0	0	42	43	20	895
Yea	4	0	0	0	43	39	918
1-5:	5	0	0	0	0	49	951

Released in year 1, shot in year 4...

V1-V5: Year of recovery

V6: never reencountered

Ф	*	V1 [‡]	V2 [‡]	V3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
release	1	59	34	27	19	8	853
	2	0	51	27	24	13	885
Year of	3	0	0	42	43	20	895
Yea	4	0	0	0	43	39	918
1-5:	5	0	0	0	0	49	951

You have to survive 3 years, and be shot in year 4 ($S_1S_2S_3f_4$)

$$P = \begin{bmatrix} f_1 & S_1 f_2 & S_1 S_2 f_3 & S_1 S_2 S_3 f_4 & S_1 S_2 S_3 S_4 f_5 & 1 - \sum_{j=1}^{n} P_{1,1:5} \\ 0 & f_2 & S_2 f_3 & S_2 S_3 f_4 & S_2 S_3 S_4 f_5 & 1 - \sum_{j=1}^{n} P_{2,1:5} \\ 0 & 0 & f_3 & S_3 f_4 & S_3 S_4 f_5 & 1 - \sum_{j=1}^{n} P_{3,1:5} \\ 0 & 0 & 0 & f_4 & S_4 f_5 & 1 - \sum_{j=1}^{n} P_{4,1:5} \\ 0 & 0 & 0 & 0 & f_5 & 1 - \sum_{j=1}^{n} P_{5,1:5} \end{bmatrix}$$

Released in year 1, shot in year 4...

V1-V5: Year of recovery V6: never reencountered

lease	
<u>a</u>	
of	
fear	
-5:	

*	V1 [‡]	V2 [‡]	V 3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1	59	34	27	19	8	853
2	0	51	27	24	13	885
3	0	0	42	43	20	895
4	0	0	0	43	39	918
5	0	0	0	0	49	951

V1-V5: Year of recovery

V6: never reencountered

853

885

895

918

951

Ф	*	V1 [‡]	V2 [‡]	V 3 [‡]	V4 [‡]	V 5 [‡]
release	- 1	59	34	27	19	8
	2	0	51	27	24	13
Year of	3	0	0	42	43	20
Yea	4	0	0	0	43	39
- 5:	5	0	0	0	0	49
~						

You have to survive 3 years, and be shot in year 5 ($S_2S_3S_4f_5$)

$$P = \begin{bmatrix} f_1 & S_1 f_2 & S_1 S_2 f_3 & S_1 S_2 S_3 f_4 & S_1 S_2 S_3 S_4 f_5 & 1 - \sum P_{1,1:5} \\ 0 & f_2 & S_2 f_3 & S_2 S_3 f_4 & S_2 S_3 S_4 f_5 & 1 - \sum P_{2,1:5} \\ 0 & 0 & f_3 & S_3 f_4 & S_3 S_4 f_5 & 1 - \sum P_{3,1:5} \\ 0 & 0 & 0 & f_4 & S_4 f_5 & 1 - \sum P_{4,1:5} \\ 0 & 0 & 0 & 0 & f_5 & 1 - \sum P_{5,1:5} \end{bmatrix}$$

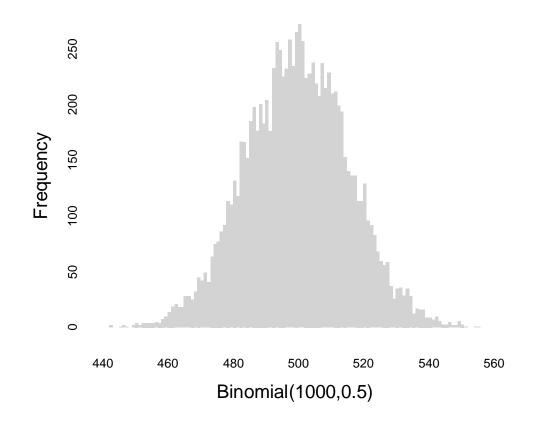
Released in year 2, shot in year 5...

How do we model that?

The binomial distribution: flip a coin 1000 times

$$y \sim \text{binomial}(\mathbf{1000}, \theta)$$

 $\theta = 0.5$





Binomial outcomes for band-recovery models

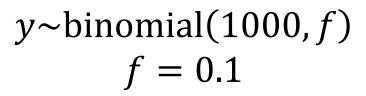
How many heads from 1000 flips?

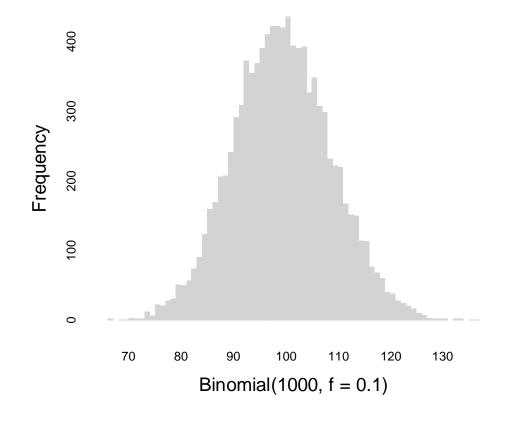
How many direct recoveries from 1000 releases?





The binomial distribution: band 1000 mallards, recover 10%





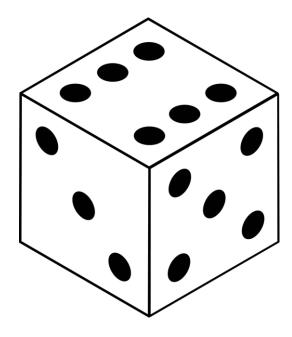




The multinomial distribution: roll a die 1000 times

 $y \sim \text{multinomial}(\mathbf{1000}, \boldsymbol{\theta})$

$$\boldsymbol{\theta} = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ \overline{6}, \overline{6}, \overline{6}, \overline{6}, \overline{6}, \overline{6} \end{bmatrix}$$



The multinomial distribution: roll a die 1000 times

$$\mathbf{y} \sim \text{multinomial}(1000, \boldsymbol{\theta})$$

$$\boldsymbol{\theta} = \left[\frac{1}{20}, \frac{1}{$$

```
> y <- rmultinom(1, 1000, rep(1/20, 20))
> as.vector(y)
[1] 47 59 59 42 47 53 49 53 50 49 54 53 47 42 53 59 52 37 53 42
> 1:20
[1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
```



Multinomial outcomes for band-recovery models

Did we roll a 1, or a 2, or a 3, or a 4,, or a 20



Shot in year 1, or shot in year 2, ... or shot in year 15, or never shot



The multinomial distribution: roll a die 1000 times

$$\boldsymbol{\theta} = \begin{bmatrix} \frac{1}{20}, \frac{1}{$$



For released ducks, the probabilities are different

$$m_1$$
~multinomial(1000, p_1)
 $p_1 = [f_1, S_1 f_2, 1 - (f_1 + S_1 f_2)]$

$$R = \begin{bmatrix} 1000 \\ 1000 \end{bmatrix}$$

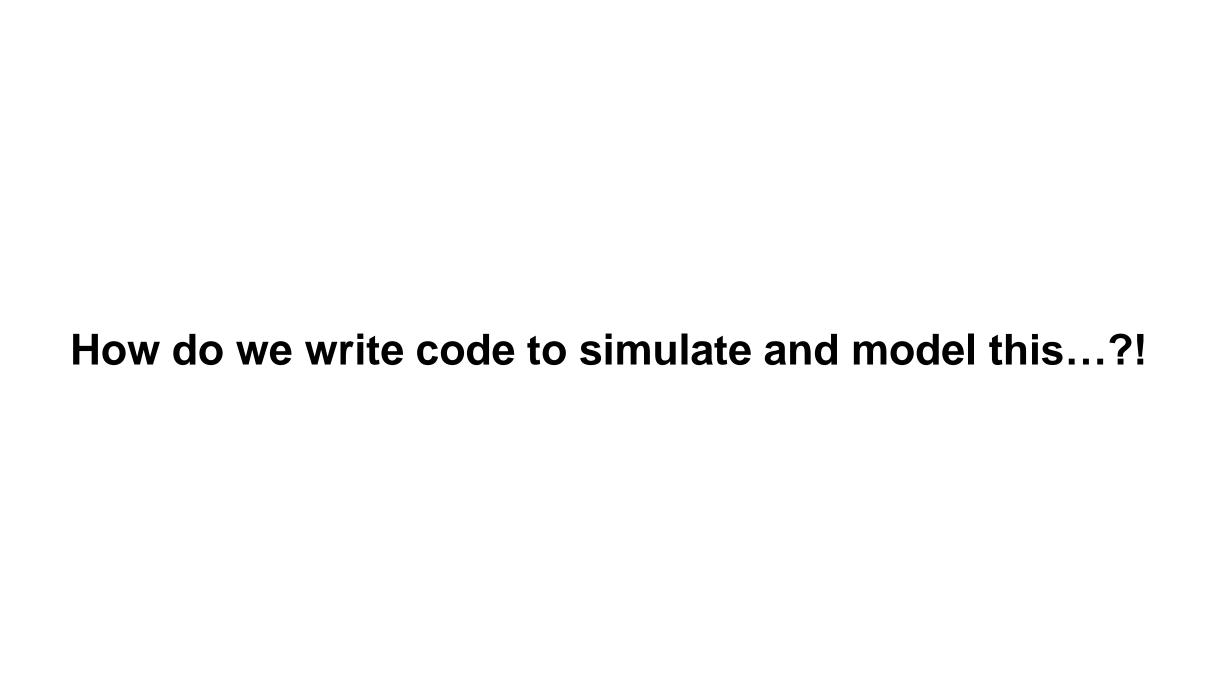
$$M = \begin{bmatrix} 100 & 50 & 850 \\ 0 & 100 & 900 \end{bmatrix}$$

For released ducks, the probabilities are different

$$m_2$$
~multinomial(1000, p_2)
 $p_2 = [0, f_2, 1 - (f_2)]$

$$R = \begin{bmatrix} 1000 \\ \mathbf{1000} \end{bmatrix}$$

$$M = \begin{bmatrix} 100 & 50 & 850 \\ \mathbf{0} & \mathbf{100} & \mathbf{900} \end{bmatrix}$$



For-loops

```
> for (i in 1:10){
  print(i)
```

For-loops

```
> for (i in 1:10){
   print(i+3)
    10
```

For-loops

```
> x <- rep(NA, 10)
> x
   [1] NA NA NA NA NA NA NA NA NA
> for (i in 1:10){
+   x[i] <- paste0('f',i)
+ }
> x
   [1] "f1" "f2" "f3" "f4" "f5" "f6" "f7" "f8" "f9" "f10"
> |
```

These are our cell probabilities

$$P = \begin{bmatrix} f_1 & S_1 f_2 & S_1 S_2 f_3 & S_1 S_2 S_3 f_4 & S_1 S_2 S_3 S_4 f_5 & 1 - \sum P_{1,1:5} \\ 0 & f_2 & S_2 f_3 & S_2 S_3 f_4 & S_2 S_3 S_4 f_5 & 1 - \sum P_{2,1:5} \\ 0 & 0 & f_3 & S_3 f_4 & S_3 S_4 f_5 & 1 - \sum P_{3,1:5} \\ 0 & 0 & 0 & f_4 & S_4 f_5 & 1 - \sum P_{4,1:5} \\ 0 & 0 & 0 & 0 & f_5 & 1 - \sum P_{5,1:5} \end{bmatrix}$$

How could I write this in code using loops?

How could I write this in code?

$$P = \begin{bmatrix} f_1 & S_1 f_2 & S_1 S_2 f_3 & S_1 S_2 S_3 f_4 & S_1 S_2 S_3 S_4 f_5 & 1 - \sum_{1} P_{1,1:5} \\ 0 & f_2 & S_2 f_3 & S_2 S_3 f_4 & S_2 S_3 S_4 f_5 & 1 - \sum_{1} P_{2,1:5} \\ 0 & 0 & f_3 & S_3 f_4 & S_3 S_4 f_5 & 1 - \sum_{1} P_{3,1:5} \\ 0 & 0 & 0 & f_4 & S_4 f_5 & 1 - \sum_{1} P_{4,1:5} \\ 0 & 0 & 0 & 0 & f_5 & 1 - \sum_{1} P_{5,1:5} \end{bmatrix}$$

Let's look for patterns

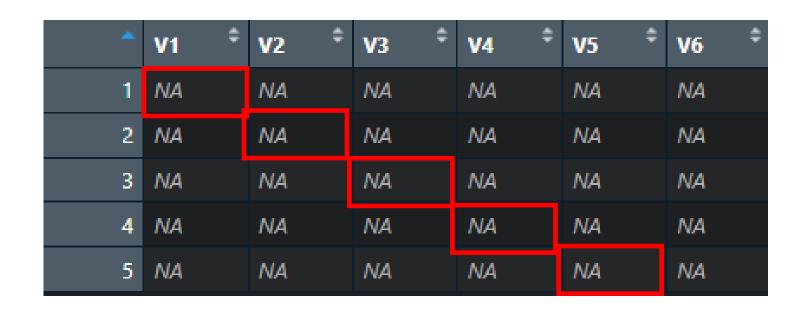
$$\begin{pmatrix}
f_1 \\
S_1 f_2 \\
S_1 S_2 f_3
\end{pmatrix} S_1 S_2 S_3 f_4 S_1 S_2 S_3 S_4 f_5 1 - \sum_{j=1}^{n} P_{1,1:5} \\
0 \begin{cases}
f_2 \\
0
\end{cases} S_2 f_3 S_2 S_3 f_4 S_2 S_3 S_4 f_5 1 - \sum_{j=1}^{n} P_{2,1:5} \\
0 \begin{cases}
0 \\
0
\end{cases} S_3 f_4 S_3 S_4 f_5 1 - \sum_{j=1}^{n} P_{3,1:5} \\
0 \begin{cases}
0 \\
0
\end{cases} 0 \begin{cases}
f_4 \\
0
\end{cases} S_4 f_5 1 - \sum_{j=1}^{n} P_{4,1:5} \\
0 \begin{cases}
0 \\
0
\end{cases} 0 \begin{cases}
0 \\
0
\end{cases} 0 \begin{cases}
0 \\
0
\end{cases} 0 \begin{cases}
f_5 \\
1 - \sum_{j=1}^{n} P_{5,1:5} \\
0 \end{cases}$$

$$p[1,1] = f_1$$

 $p[2,2] = f_2$
 $p[3,3] = f_3$
 $p[4,4] = f_4$
 $p[5,5] = f_5$

Main diagonal

Constructing cell probabilities (a five-year study)

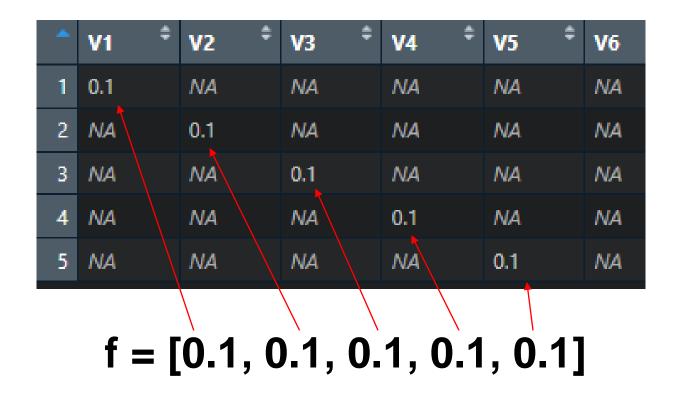


$$f = [0.1, 0.1, 0.1, 0.1, 0.1]$$

```
for (t in 1:nT){
  p[t,t] <- f[t]
}</pre>
```

$$p[1,1] = f_1$$

 $p[2,2] = f_2$
 $p[3,3] = f_3$
 $p[4,4] = f_4$
 $p[5,5] = f_5$



$$P = \begin{bmatrix} f_1 & S_1 f_2 & S_1 S_2 f_3 & S_1 S_2 S_3 f_4 & S_1 S_2 S_3 S_4 f_5 & 1 - \sum_{j=0}^{n} P_{1,1:5} \\ 0 & f_2 & S_2 f_3 & S_2 S_3 f_4 & S_2 S_3 S_4 f_5 & 1 - \sum_{j=0}^{n} P_{2,1:5} \\ 0 & 0 & f_3 & S_3 f_4 & S_3 S_4 f_5 & 1 - \sum_{j=0}^{n} P_{3,1:5} \\ 0 & 0 & f_4 & S_4 f_5 & 1 - \sum_{j=0}^{n} P_{4,1:5} \\ 0 & 0 & 0 & f_5 & 1 - \sum_{j=0}^{n} P_{4,1:5} \\ 0 & 0 & 0 & f_5 & 1 - \sum_{j=0}^{n} P_{5,1:5} \end{bmatrix}$$

$$p[2,1] = 0$$

$$p[3,2] = 0$$

$$p[4,2] = 0$$

$$p[4,3] = 0$$

$$p[5,1] = 0$$

$$p[5,2] = 0$$

$$p[5,3] = 0$$

$$p[5,3] = 0$$

$$p[5,4] = 0$$

Below the main diagonal = 0

```
for (t in 2:nT){
  for (j in 1:(t-1)){
    p[t,j] <- 0
  }
}</pre>
```

*	V1 [‡]	V2 [‡]	V 3 [‡]	V4 [‡]	V 5 [‡]	V6
1	0.1	NA	NA	NA	NA	NA
2	0.0	0.1	NA	NA	NA	NA
3	0.0	0.0	0.1	NA	NA	NA
4	0.0	0.0	0.0	0.1	NA	NA
5	0.0	0.0	0.0	0.0	0.1	NA

$$t = 2$$

```
for (t in 2:nT){
  for (j in 1:(t-1)){
    p[t,j] <- 0
  }
}</pre>
```

*	V1 [‡]	V 2 [‡]	V 3 [‡]	V4 [‡]	V5 [‡]	V6
1	0.1	NA	NA	NA	NA	NA
2	0.0	0.1	NA	NA	NA	NA
3	0.0	0.0	0.1	NA	NA	NA
4	0.0	0.0	0.0	0.1	NA	NA
5	0.0	0.0	0.0	0.0	0.1	NA

$$t = 3$$

$$p[2,1] = 0$$

$$p[3,1] = 0$$

$$p[3,2] = 0$$

$$p[4,1] = 0$$

$$p[4,2] = 0$$

$$p[5,1] = 0$$

$$p[5,1] = 0$$

$$p[5,2] = 0$$

$$p[5,3] = 0$$

$$p[5,4] = 0$$

```
for (t in 2:nT){
  for (j in 1:(t-1)){
    p[t,j] <- 0
  }
}</pre>
```

^	V1 [‡]	V2 [‡]	V3 [‡]	V4 [‡]	V 5 [‡]	V 6
1	0.1	NA	NA	NA	NA	NA
2	0.0	0.1	NA	NA	NA	NA
3	0.0	0.0	0.1	NA	NA	NA
4	0.0	0.0	0.0	0.1	NA	NA
5	0.0	0.0	0.0	0.0	0.1	NA

```
for (t in 2:nT){
  for (j in 1:(t-1)){
    p[t,j] <- 0
  }
}</pre>
```

•	V1 [‡]	V 2 [‡]	V 3 [‡]	V4 [‡]	V5 [‡]	V6
1	0.1	NA	NA	NA	NA	NA
2	0.0	0.1	NA	NA	NA	NA
3	0.0	0.0	0.1	NA	NA	NA
4	0.0	0.0	0.0	0.1	NA	NA
5	0.0	0.0	0.0	0.0	0.1	NA

$$P = \begin{bmatrix} f_1 & S_1 f_2 & S_1 S_2 f_3 & S_1 S_2 S_3 f_4 & S_1 S_2 S_3 S_4 f_5 \\ 0 & f_2 & S_2 f_3 & S_2 S_3 f_4 & S_2 S_3 S_4 f_5 \\ 0 & 0 & f_3 & S_3 f_4 & S_3 S_4 f_5 \\ 0 & 0 & 0 & f_4 & S_4 f_5 \\ 0 & 0 & 0 & 0 & f_5 & 1 - \sum_{A_{1:5}} P_{1:5} \end{bmatrix} \begin{bmatrix} p[1,2] = S_1 f_2 \\ p[1,3] = S_1 S_2 f_3 \\ p[1,4] = \dots \\ p[1,5] = \dots \\ p[2,3] = S_2 f_3 \\ p[2,4] = S_2 S_3 f_4 \\ p[2,5] = \dots \\ p[3,4] = S_3 f_4 \\ p[3,5] = S_3 f_4 \\ p[4,5] = S_4 f_5 \end{bmatrix}$$

$$p[1,2] = S_1f_2$$

 $p[1,3] = S_1S_2f_3$
 $p[1,4] = ...$
 $p[1,5] = ...$
 $p[2,3] = S_2f_3$
 $p[2,4] = S_2S_3f_4$
 $p[2,5] = ...$
 $p[3,4] = S_3f_4$
 $p[3,5] = S_3S_4f_5$
 $p[4,5] = S_1f_4$

```
for (t in 1:(nT-1)){
   for (j in (t+1):nT){
     p[t,j] <- prod(S[t:(j-1)]) * f[j]
   }
}</pre>
```

*	V1 [‡]	V2 [‡]	V3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1	0.1	0.07	0.049	0.0343	0.02401	NA
2	0.0	0.10	0.070	0.0490	0.03430	NA
3	0.0	0.00	0.100	0.0700	0.04900	NA
4	0.0	0.00	0.000	0.1000	0.07000	NA
5	0.0	0.00	0.000	0.0000	0.10000	NA

$$f = [0.1, 0.1, 0.1, 0.1, 0.1]$$

 $S = [0.7, 0.7, 0.7, 0.7]$

$$p[1,2] = S_1f_2$$

 $p[1,3] = S_1S_2f_3$

$$P = \begin{bmatrix} f_1 & S_1 f_2 & S_1 S_2 f_3 & S_1 S_2 S_3 f_4 & S_1 S_2 S_3 S_4 f_5 & 1 - \sum_{1} P_{1,1:5} \\ 0 & f_2 & S_2 f_3 & S_2 S_3 f_4 & S_2 S_3 S_4 f_5 & 1 - \sum_{1} P_{2,1:5} \\ 0 & 0 & f_3 & S_3 f_4 & S_3 S_4 f_5 & 1 - \sum_{1} P_{3,1:5} \\ 0 & 0 & 0 & f_4 & S_4 f_5 & 1 - \sum_{1} P_{4,1:5} \\ 0 & 0 & 0 & 0 & f_5 & 1 - \sum_{1} P_{5,1:5} \end{bmatrix}$$

```
for (t in 1:nT){
  p[t,nT+1] <- 1 - sum(m[t,1:nT])
}</pre>
```

*	V1 [‡]	V2 [‡]	V3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1	0.1	0.07	0.049	0.0343	0.02401	0.72269
2	0.0	0.10	0.070	0.0490	0.03430	0.74670
3	0.0	0.00	0.100	0.0700	0.04900	0.78100
4	0.0	0.00	0.000	0.1000	0.07000	0.83000
5	0.0	0.00	0.000	0.0000	0.10000	0.90000

Simulating data

*	V1 [‡]	V2 ‡	V 3 ‡	V4 [‡]	V5 [‡]	V6 [‡]
1	0.1	0.07	0.049	0.0343	0.02401	0.72269
2	0.0	0.10	0.070	0.0490	0.03430	0.74670
3	0.0	0.00	0.100	0.0700	0.04900	0.78100
4	0.0	0.00	0.000	0.1000	0.07000	0.83000
5	0.0	0.00	0.000	0.0000	0.10000	0.90000

The multinomial distribution: roll a die 1000 times

 $\mathbf{y} \sim \text{multinomial}(1000, \boldsymbol{\theta})$ $\boldsymbol{\theta} = \left[\frac{1}{20}, \frac{1}{$

```
> y <- rmultinom(1, 1000, rep(1/20, 20))
> as.vector(y)
[1] 47 59 59 42 47 53 49 53 50 49 54 53 47 42 53 59 52 37 53 42
> 1:20
[1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
```



 m_1 ~multinomial(1000, p_1)

*	V1 [‡]	V2 [‡]	V 3 [‡]	V4 [‡]	V 5 [‡]	V6 [‡]
1	0.1	0.07	0.049	0.0343	0.02401	0.72269
2	0.0	0.10	0.070	0.0490	0.03430	0.74670
3	0.0	0.00	0.100	0.0700	0.04900	0.78100
4	0.0	0.00	0.000	0.1000	0.07000	0.83000
5	0.0	0.00	0.000	0.0000	0.10000	0.90000

 m_2 ~multinomial(1000, p_2)

*	V1 [‡]	V2 ‡	V 3 [‡]	V4 [‡]	V5 [‡]	V6 [‡]
1	0.1	0.07	0.049	0.0343	0.02401	0.72269
2	0.0	0.10	0.070	0.0490	0.03430	0.74670
3	0.0	0.00	0.100	0.0700	0.04900	0.78100
4	0.0	0.00	0.000	0.1000	0.07000	0.83000
5	0.0	0.00	0.000	0.0000	0.10000	0.90000

```
for (t in 1:nT){
   m[t,1:(nT+1)] <- rmultinom(1, 1000, p[t,1:(nT+1)])
}</pre>
```

•	V1 [‡]	V 2 [‡]	V 3 ‡	V4 [‡]	V 5 [‡]	V6 [‡]
1	88	72	39	35	25	741
2	0	106	66	47	41	740
3	0	0	85	90	53	772
4	0	0	0	115	67	818
5	0	0	0	0	88	912