

Estimating Abundance from Camera Data



Introductions

Schedule / Format

- Today
 - Morning
 - Understand the camera problem
 - Sampling
 - Introduce estimators
 - Afternoon
 - Estimation details
 - Assumptions
 - Field considerations

Schedule / Format

- Tomorrow
 - Morning
 - Participants work on their own data
- Please ask questions
 - We can modify the schedule to meet your needs

Motivating the Camera Problem

Camera data are unique!

Camera Observations

- Photo
 - Is there an animal in the photo?
 - What kind of animal (species/age/sex)?
 - Where was the camera?
 - When was the photo taken?
 - Is the animal marked in some way?
 - Can you read the mark with certainty?

Common Analysis Methods

- Occupancy
 - Capture-recapture
 - Spatial capture-recapture
 - N -mixture
-
- All miss fundamental issues with camera data

Issues with Camera Data

- Movement

- Animals must move to get into the field of view
- Trigger motion sensor



Occupancy



Capture-recapture



Spatial capture-recapture



N-mixture

Issues with Camera Data

- Logistical complexity
 - Cameras are expensive
 - Deployment is complex
 - Maintenance is complicated
 - Number of photos is huge
 - Animal populations operate at large spatial scales



Occupancy



Capture-recapture



Spatial capture-recapture



N-mixture

Issues with Camera Data

- Individual identification

- Animals can be difficult to enumerate in a photo
- Individual ID may not be possible



Occupancy



Capture-recapture



Spatial capture-recapture



N-mixture

Objective

- Estimator for camera data
 - Not affected by movement
 - Does not require marked animals
 - Does not require counts of animals in a photo
 - Logistically feasible at large scales
 - Estimates abundance

Sampling

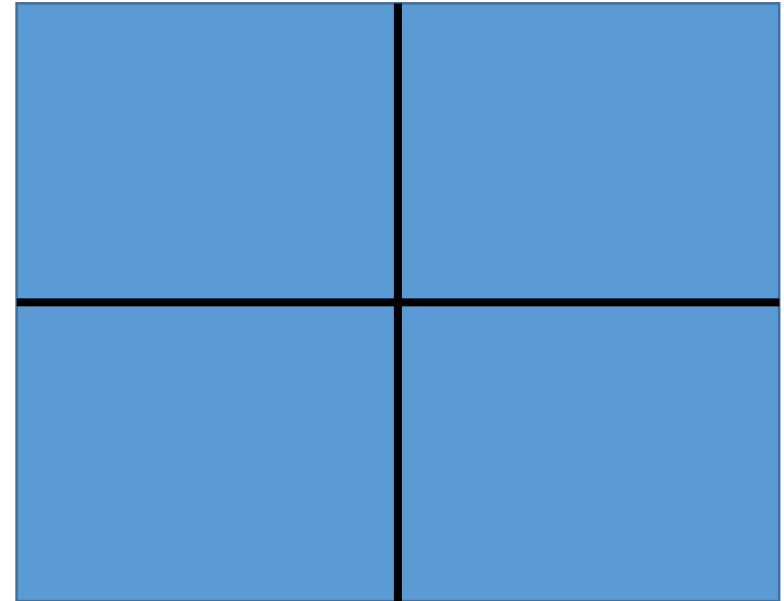
The answer is hidden in plain sight

Sampling

- Basics are common knowledge
 - Use a subset to make inference to the whole
- Details are not intuitive

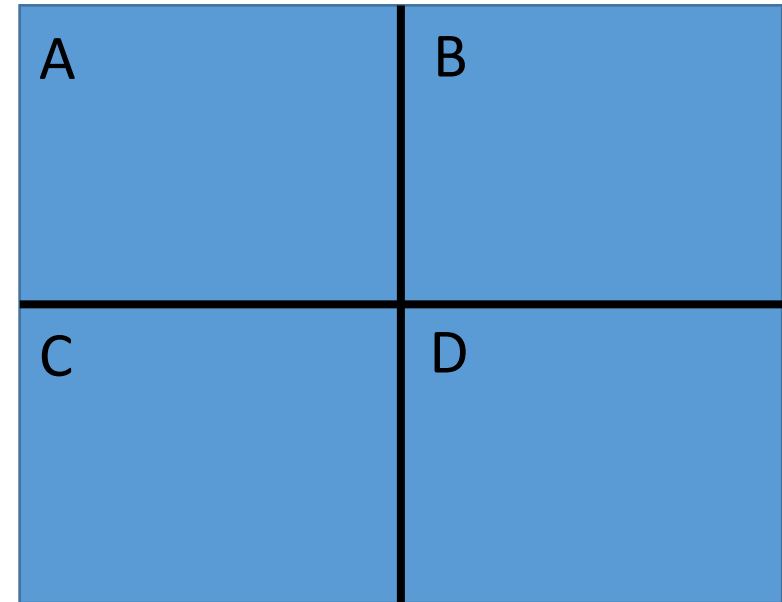
Motivating Example

- Sampling frame (entire blue area)
 - Area of inference
 - Study area
- Sampling Units (smaller squares)
 - Area measured
 - Camera viewshed
 - Survey plot



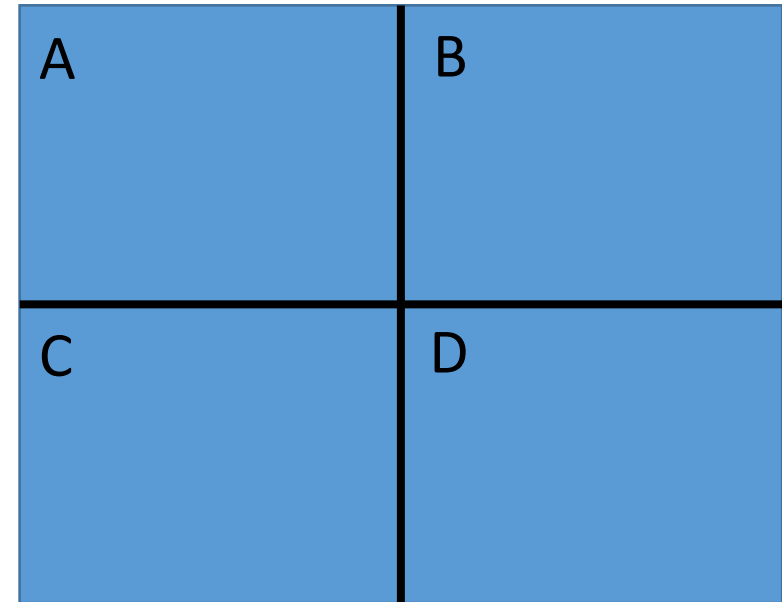
Motivating Example

- Sample
 - Selected sampling units for inference
- Sampling distribution
 - All possible combinations of samples of the same size
 - How many combinations of samples of size 1 are there?
 - How many of size 2?



Motivating Example

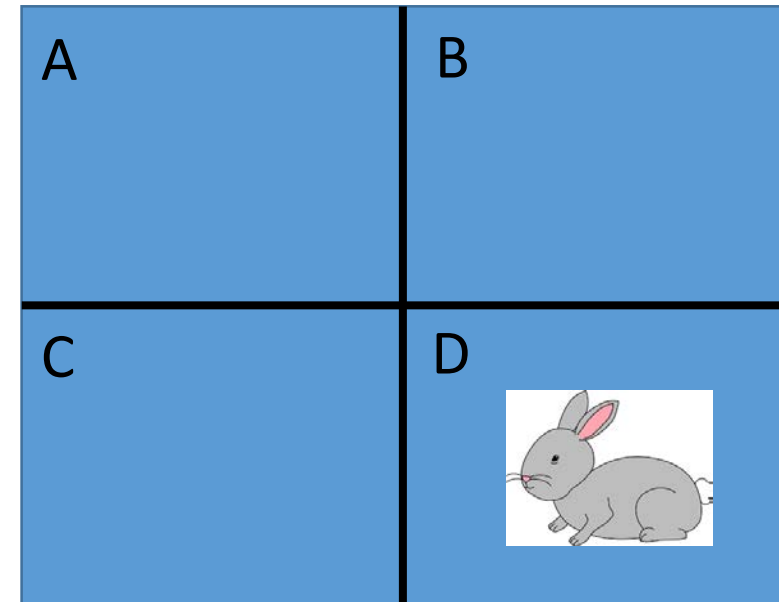
- Sample of size 1
 - Represents $\frac{1}{4}$ of sampling frame
 - Estimate = Count / $(1/4)$



Motivating Example

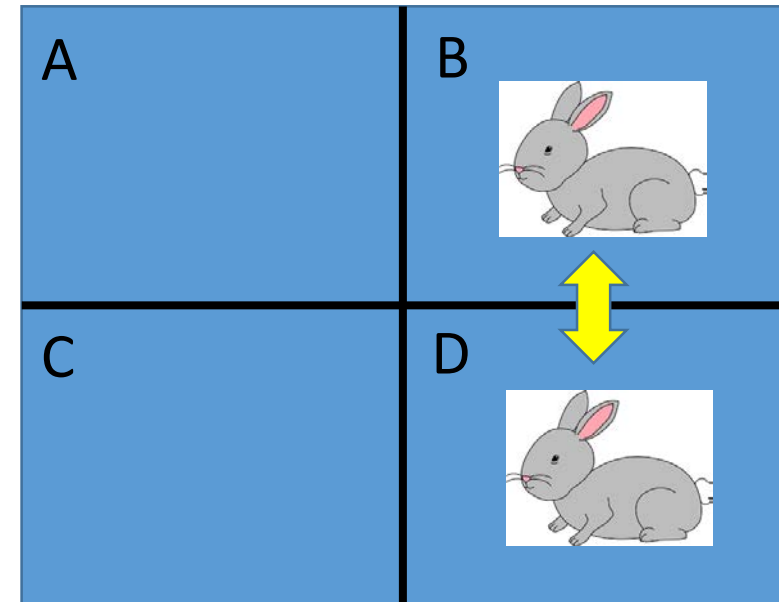
- Sample of size 1
 - Represents $\frac{1}{4}$ of sampling frame
 - Estimate = Count / $(1/4)$

Plot	Count	Fraction	Estimate
A	0	0.25	0
B	0	0.25	0
C	0	0.25	0
D	1	0.25	4
Average			1



Motivating Example – now add movement

- Sample of size 1
 - Represents $\frac{1}{4}$ of sampling frame
 - Estimate = Count / $(1/4)$

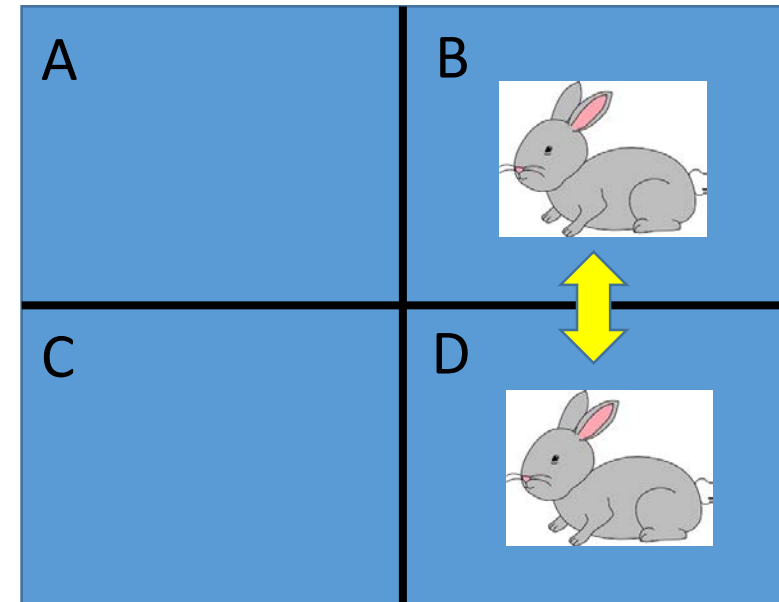


Motivating Example – now add movement

- Sample of size 1
 - Represents $\frac{1}{4}$ of sampling frame
 - Estimate = Count / $(1/4)$

Plot	Count	Fraction	Estimate
A	0	0.25	0
B	1	0.25	4
C	0	0.25	0
D	1	0.25	4
Average			2

!!!

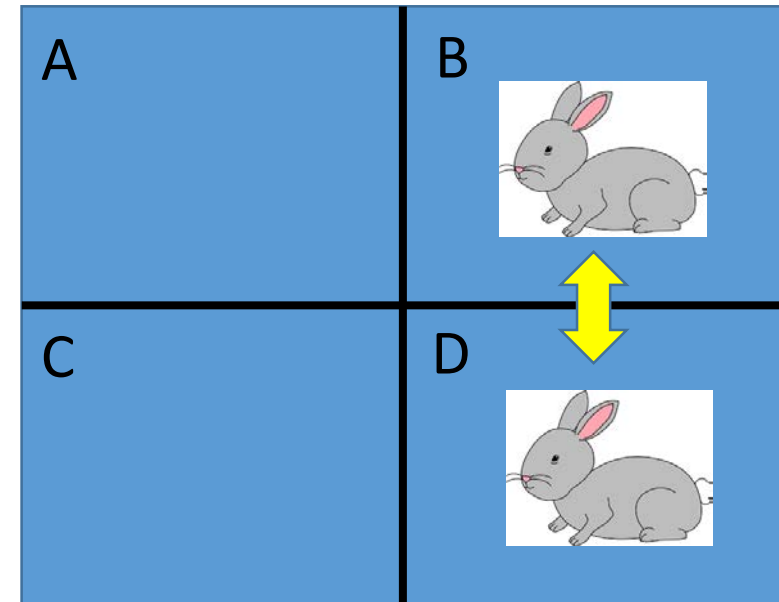


Motivating Example – now add movement

- Bias – how did it happen?
 - No double counting
 - No plots repeated across samples

Plot	Count	Fraction	Estimate
A	0	0.25	0
B	1	0.25	4
C	0	0.25	0
D	1	0.25	4
Average			2

!!!



Sampling – take home messages

- Sampling provides unbiased estimates of anything you can measure without a model
- Movement of the object of interest (animal) must be slow relative to the duration of the survey

Camera data estimators

Data Generation

- Photo
 - Time photo was taken
 - Area visible in the photo
 - Animal present or not
 - Camera with the detection
 - Time Lapse or motion trigger

Data Generation

- When will we get a picture of a cougar?



Data Generation

- Distributions
 - What is the distribution of times to the next photo?

Data Generation

- Distributions
 - What is the distribution of times to the next photo?
 - Exponential

Data Generation

- Distributions
 - What is the distribution of times to the next photo?
 - Exponential
- Let's step through the distribution on the whiteboard...



“Events”

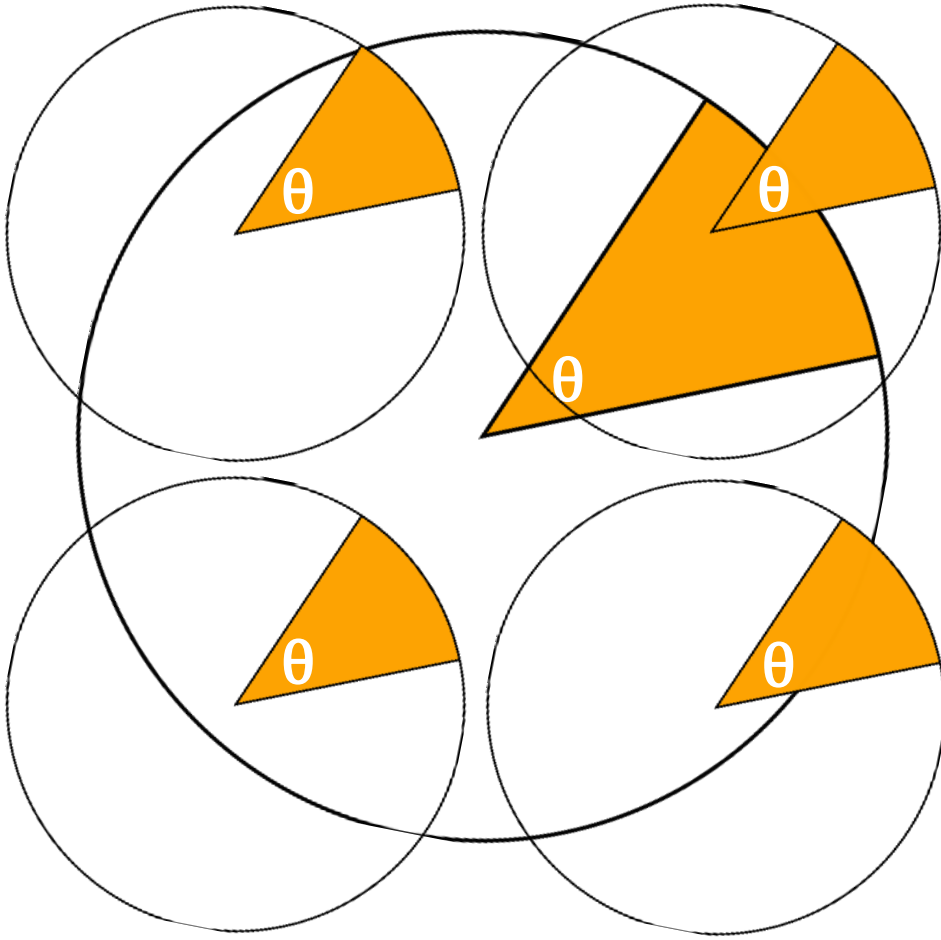


$$N \sim \text{Pois}(\lambda)$$
$$S \sim \text{Exp}(\lambda)$$

“Waiting time”



Cameras



$$\hat{N} = \frac{\text{total counted}}{\# \text{ plots sampled}} * \frac{\text{study area size}}{\pi r^2 \frac{\theta}{360}}$$



Abundance ↑

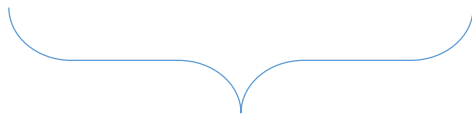
Photos ↑

Abundance ↑

Cameras with animals ↑

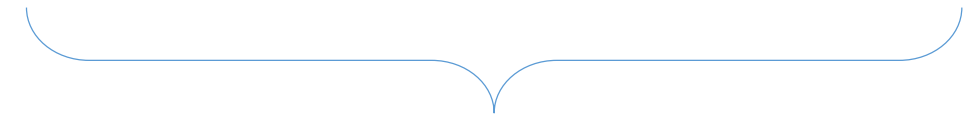
Abundance ↑

Cameras until detection ↓



Want to estimate

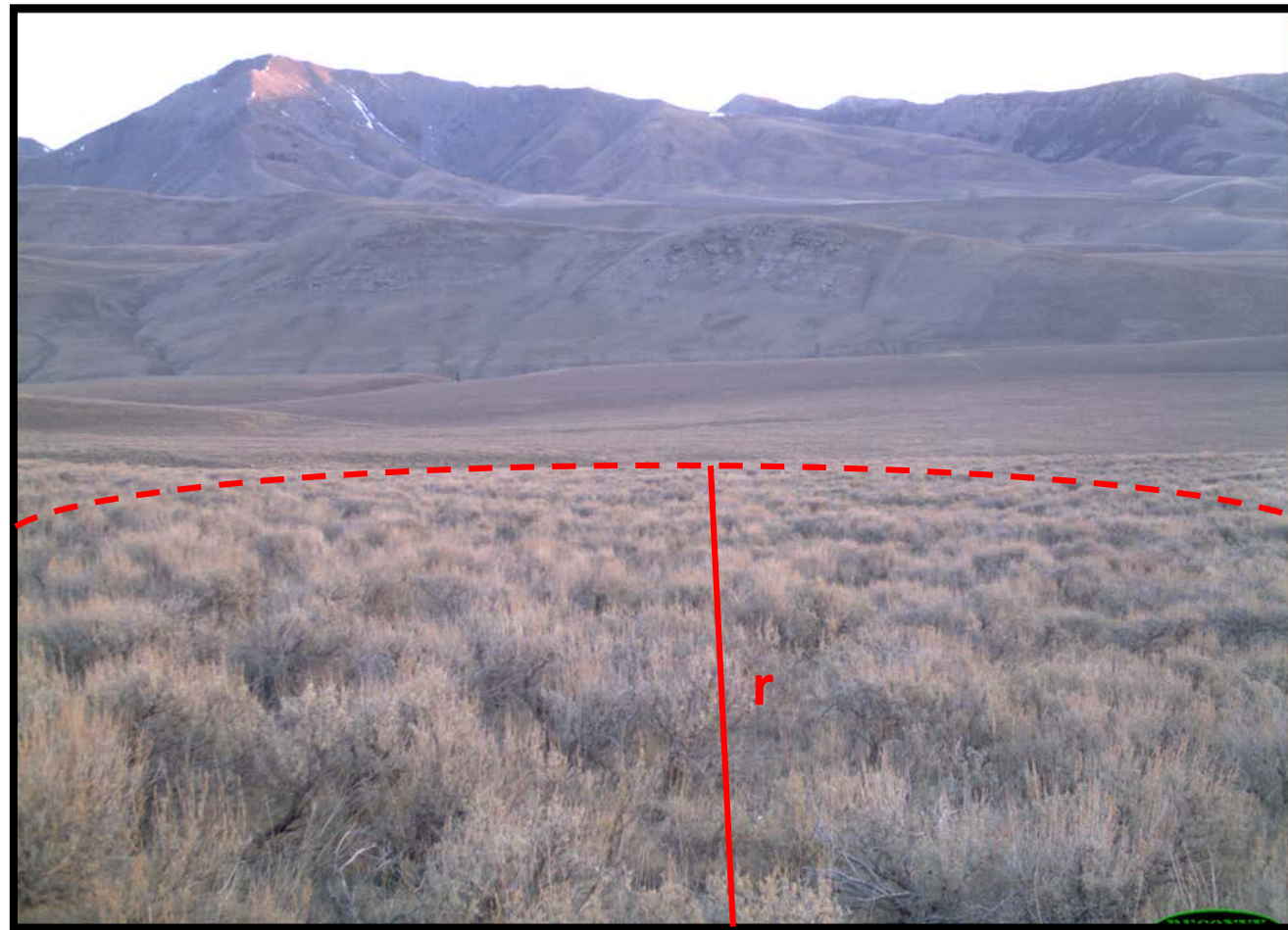
“Event”



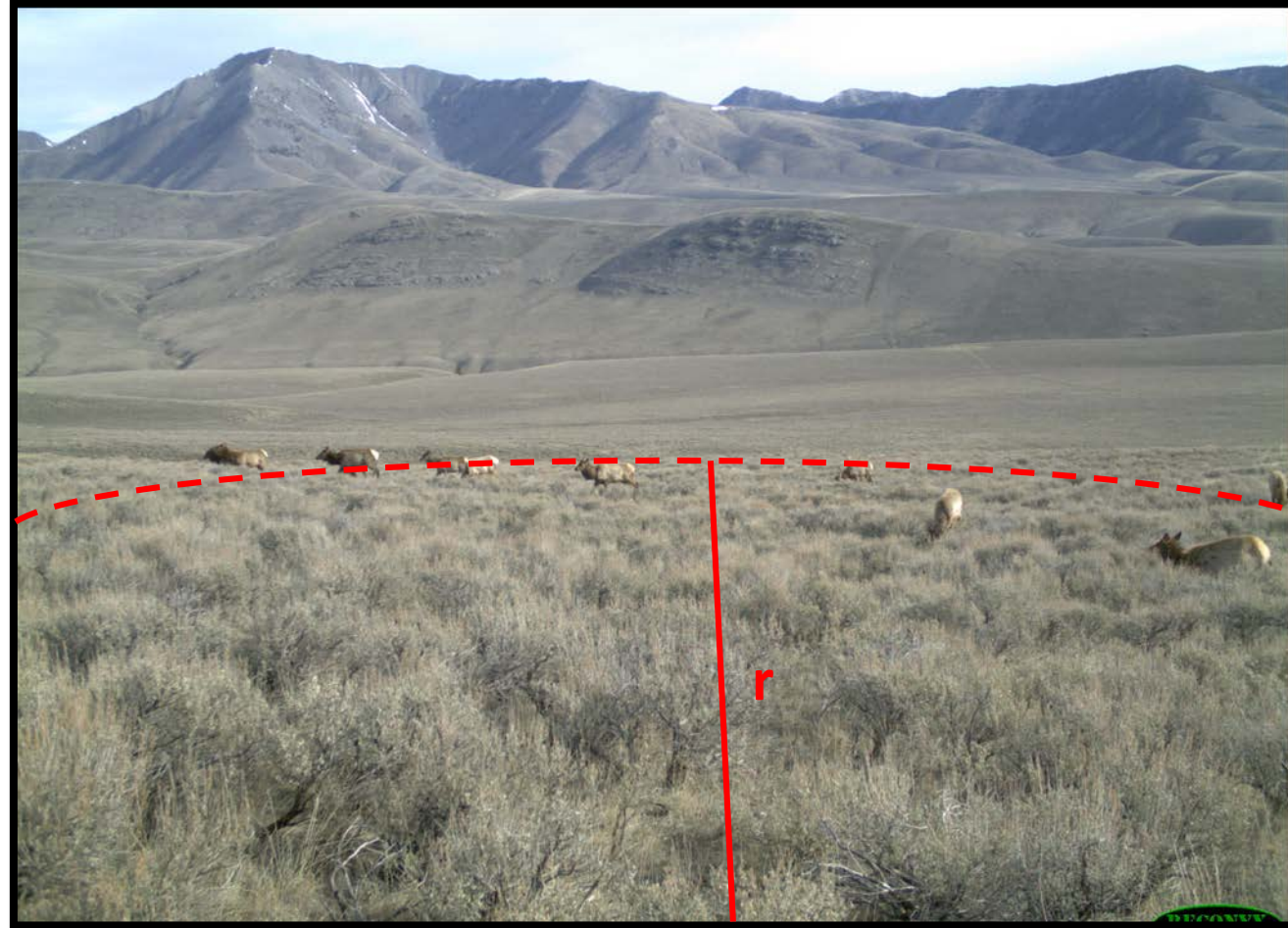
Observe

“Waiting Time”

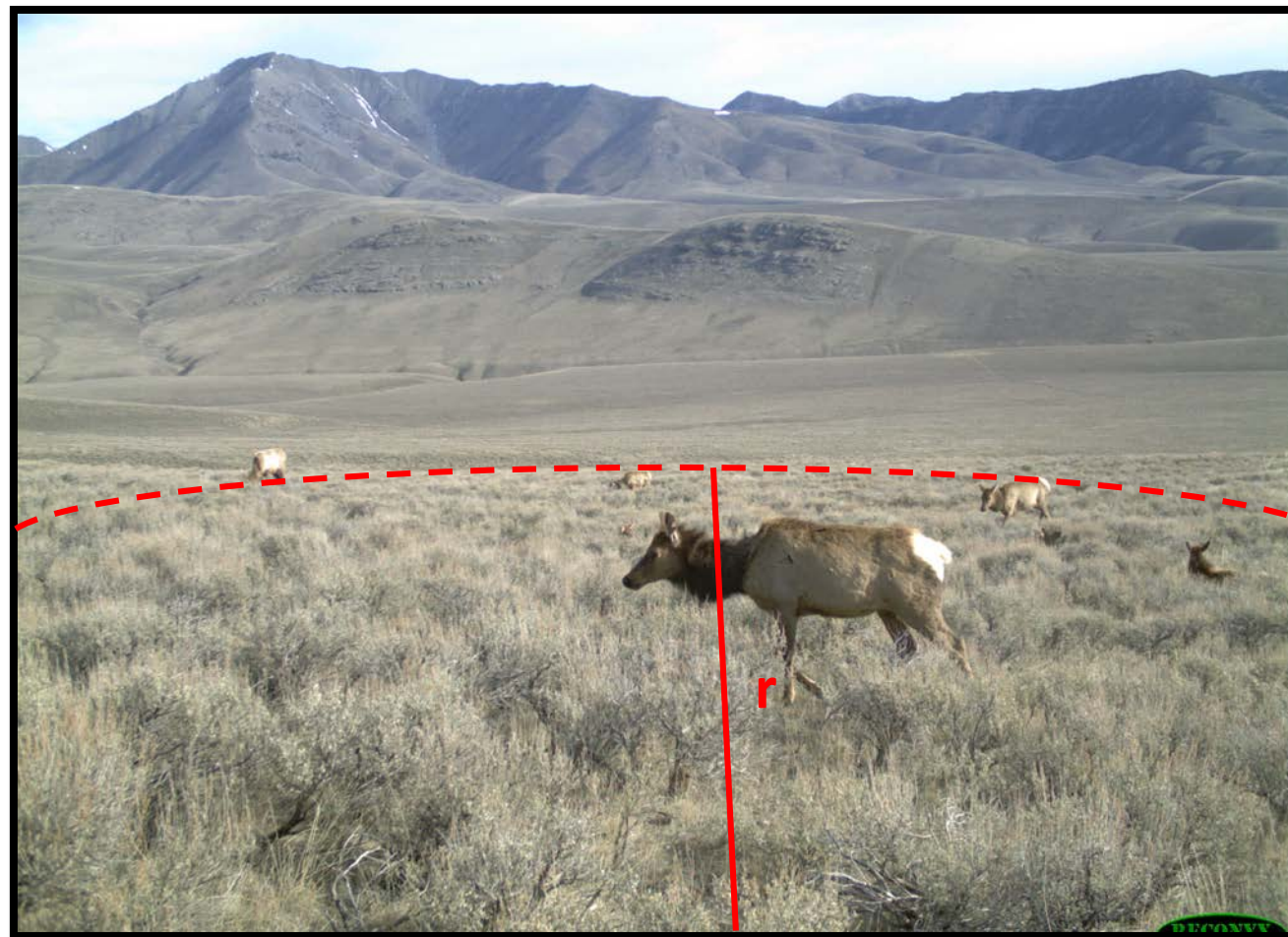
6:00 am



8:00 am



10:00 am



Space-to-Event

6:00 am



Camera 1



Camera 2



Camera 3



Camera 4



Camera 5



Camera 6

Space-to-Event Model

6:00	8:00	10:00	12:00	2:00	4:00
3					

8:00 am



Camera 1



Camera 2



Camera 3



Camera 4



Camera 5



Camera 6



Space-to-Event Model

6:00	8:00	10:00	12:00	2:00	4:00
3	2				

8:00 am



Camera 1



Camera 2



Camera 3



Camera 4



Camera 5



Camera 6



Space-to-Event Model

6:00	8:00	10:00	12:00	2:00	4:00
3	2	5	NA	1	NA

$$\mathcal{L}(\lambda|T_{ij}) = \prod_{j=1}^J \prod_{i=1}^M \sum_{z=0}^{\infty} \left(I_{(T \leq t)} p(1-p)^z \frac{\lambda_i^{z+1}}{\Gamma(z+1)} T_{ij}^z e^{-\lambda_i T_{ij}} + I_{(1-(T \leq t))} p(1-p)^z \int_t^{\infty} \frac{1}{\Gamma(z+1)} \gamma(z+1, \lambda_i T_{ij}) d\lambda \right)$$

Time-to-Event Model

8:00 am



Time-to-Event Model

8:00 am



$$T = 1$$

Time-to-Event Model

8:00 am



$$\overline{T=1}$$

Time-to-Event Model

9:00 am



$$\cancel{T = 1}$$

$$T = 2$$

Time-to-Event Model

9:00 am



$$\cancel{T=1}$$

$$\cancel{T=2}$$

Time-to-Event Model

10:00



$$\cancel{T=1}$$

$$\cancel{T=2}$$

$$T=3$$

Time-to-Event Model

10:00 am



$$\cancel{T=1}$$

$$\cancel{T=2}$$

$$T=3$$



Time-to-Event Model

	Day 1	Day 2	Day 3	Day 4	Day 5	Day 6
Cam 1	3					
Cam 2						
Cam 3						
Cam 4						



Time-to-Event Model

	Day 1	Day 2	Day 3	Day 4	Day 5	Day 6
Cam 1	3	1	NA	5	NA	NA
Cam 2	NA	NA	NA	NA	2	NA
Cam 3	NA	5	4	NA	NA	NA
Cam 4	5	NA	NA	NA	NA	4



Time-to-Event Model

	Day 1	Day 2	Day 3	Day 4	Day 5	Day 6
Cam 1	3	1	NA	5	NA	NA
Cam 2	NA	NA	NA	NA	2	NA
Cam 3	NA	5	4	NA	NA	NA
Cam 4	5	NA	NA	NA	NA	4

6:00 am



Camera 1



Camera 2



Camera 3



Camera 4



Camera 5



Camera 6

Space-to-Event Model

6:00	8:00	10:00	12:00	2:00	4:00
3					

8:00 am



Camera 1



Camera 2



Camera 3



Camera 4



Camera 5



Camera 6



Space-to-Event Model

6:00	8:00	10:00	12:00	2:00	4:00
3	2	5	NA	1	NA

$N \sim \text{Pois}(\lambda)$

$S \sim \text{Exp}(\lambda)$

Abundance ↑

Photos ↑

Abundance ↑

Time until detection ↓

Want to estimate

Observe

“Event”

“Waiting Time”

Time-to-Event Model

8:00 am



Time-to-Event Model

8:00 am



$$T = 1$$

Time-to-Event Model

8:00 am



$$\overline{T=1}$$

Time-to-Event Model

9:00 am



$$\cancel{T = 1}$$

$$T = 2$$

Time-to-Event Model

9:00 am



$$\cancel{T=1}$$

$$\cancel{T=2}$$

Time-to-Event Model

10:00



$$\cancel{T=1}$$

$$\cancel{T=2}$$

$$T=3$$

Time-to-Event Model

10:00 am



$$\cancel{T=1}$$

$$\cancel{T=2}$$

$$T=3$$



Time-to-Event Model

	Obs 1	Obs 2	Obs 3	Obs 4	Obs 5	Obs 6
Cam 1	3					
Cam 2						
Cam 3						
Cam 4						



Time-to-Event Model

	Obs 1	Obs 2	Obs 3	Obs 4	Obs 5	Obs 6
Cam 1	3	1	NA	5	NA	NA
Cam 2	NA	NA	NA	NA	2	NA
Cam 3	NA	5	4	NA	NA	NA
Cam 4	5	NA	NA	NA	NA	4



Time-to-Event Model

	Obs 1	Obs 2	Obs 3	Obs 4	Obs 5	Obs 6
Cam 1	3	1	NA	5	NA	NA
Cam 2	NA	NA	NA	NA	2	NA
Cam 3	NA	5	4	NA	NA	NA
Cam 4	5	NA	NA	NA	NA	4

$$N \sim \text{Pois}(\lambda)$$

$$T \sim \text{Exp}(\lambda)$$

Three estimators

- Instantaneous
- Space-to-Event
- Time-to-Event

More details...

Space-to-event

- Algorithm
 - Time lapse photos
 - At each photo event
 - List all active cameras
 - Randomize list order
 - Sum area covered by cameras until the first animal of interest is detected
 - Repeat for each time lapse event
 - Fit exponential distribution to the areas
 - Rate parameter is animals per unit area
 - Multiply by total sampling frame area to estimate abundance

Time-to-Event

- Algorithm
 - Motion trigger photos
 - For each camera
 - Measure time steps to next detection
 - Repeat for each camera event
 - Fit exponential distribution to the time steps
 - Rate parameter is animals per unit area
 - Multiply by total sampling frame area to estimate abundance

Instantaneous Estimator

- Algorithm
 - Time lapse photos
 - At each photo event
 - Count animals in each photo
 - Repeat for each time lapse event
 - Fit average animals per unit area
 - Multiply by total sampling frame area to estimate abundance

Assumptions

Sample Size

Potential Problems

- Kenneth

Field logistics

- Camera placement
- Numbers of cameras
- Study duration

Photo storage and processing

Future directions