

Counting Animals Part II: Sample Counts

EFB 390: Wildlife Ecology and Management

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Drawbacks of total counts / censusing

Expensive & labor-time intensive

Impractical for MOST species / systems

- need to ALL be **visible**
- the **ENTIRE** study area needs to be survey-able

Hard to assess precision



Hippos

Is the great Elephant Census a Census?

The Great Elephant Census | HHMI BioInteractive Video



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Sample counts

Simple idea:

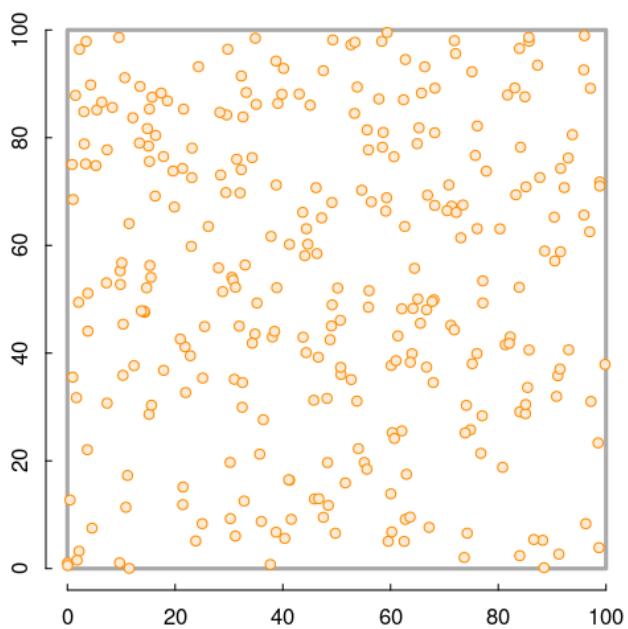
- count *some* of the individuals
- extrapolate!

In practice:

- Involves some tricky statistics and modeling!
- Necessarily - less *precise* due to *sampling error*.
- BUT ... if properly done ... more *accurate* and **much less effort**.

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A random population



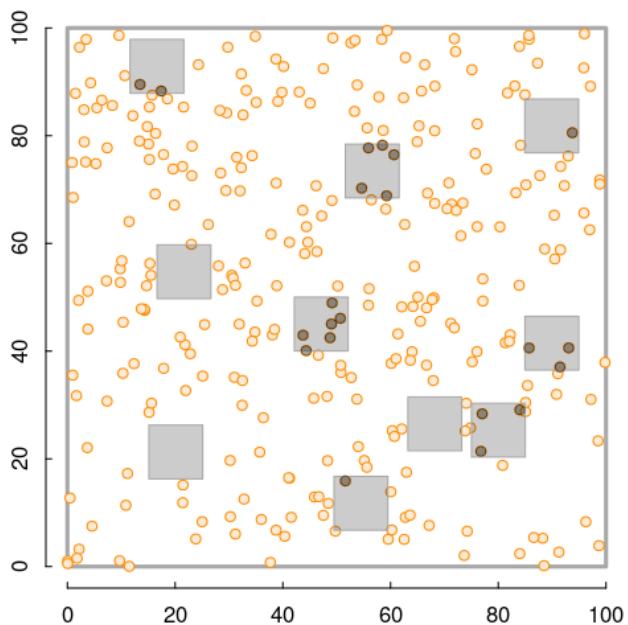
Population density

$$N = A \times D$$

- N - total count
- A - total area
- D - overall density

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Sampling from the population



Sample density:

$$n_{sample} = \sum_{i=1}^k n_i$$

$$a_{sample} = \sum_{i=1}^k a_i$$

$$d_{sample} = \frac{n_{sample}}{a_{sample}}$$

Squares, aka, quadrats

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Sample vs. Population

	Population	Sample
size	N	n_s
area	A	a_s
density	D	d_s

Note: sample density is an *estimate* of total density. So $\widehat{D} = d_s$.

True population:

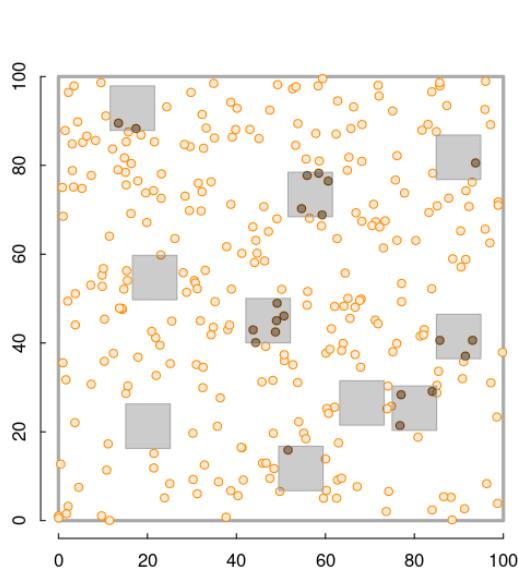
$$N = A \times D$$

Population **estimate** (best guess for N): just replace true (unknown) density D with *sampling estimate* of density d_s :

$$\widehat{N} = A \times \widehat{D} = A \times \frac{n_s}{a_s}$$

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Example



Data

10 quadrats; 10x10 km each

$n = \{0, 0, 5, 0, 3, 1, 2, 3, 6, 1\}$

note: variability / randomness!

Analysis

$$n_s = \sum n_i = 21$$

$$d_s = \widehat{D} = \frac{35}{10 \times 10 \times 10} = 0.021$$

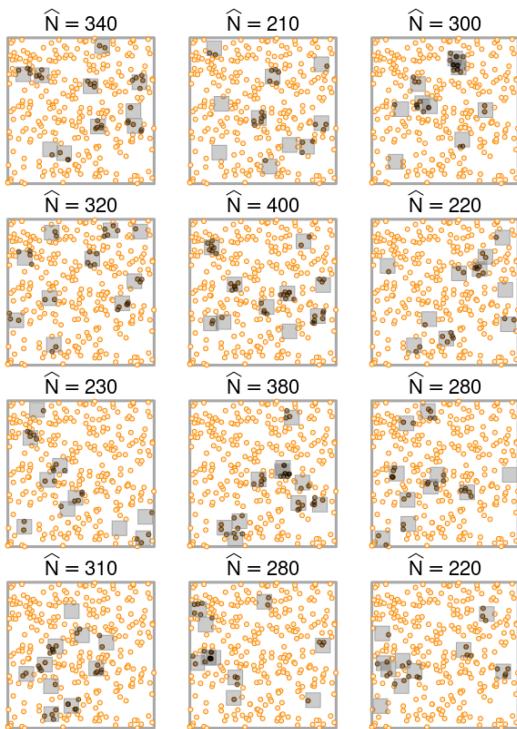
$$A = 100 \times 100$$

final estimate:

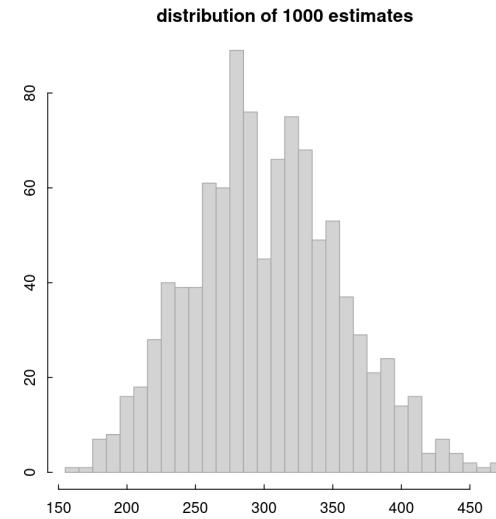
$$\widehat{N} = \widehat{D} \times A = 100 \times 100 \times 0.021 = 210$$

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What happens when we do this many times?



Every time you do this, you get a different value for \hat{N} .



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Statistics

Mean of estimates:

$$\hat{N} = 301.5$$

S.D. of estimate:

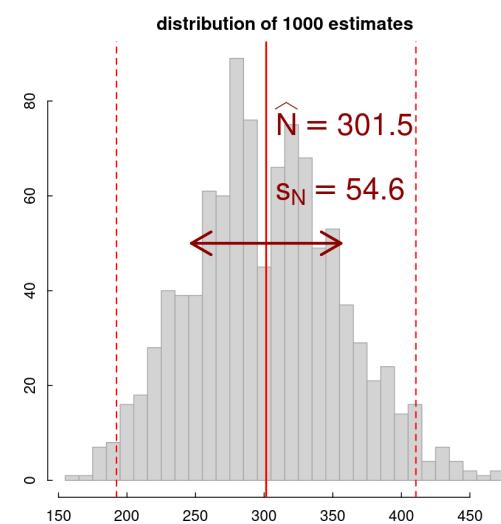
$$s_{\hat{N}} = 54.6$$

important: the *standard deviation of an estimate* = **standard error, SE**

95% Confidence Interval:

$$\hat{N} \pm 1.96 \times SE = \{195 - 408\}$$

note: the 1.96 is the number of standard deviations that captures 95% of a Normal distribution.



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General principle: The bigger the sample, the smaller the error.

1. If $a_s \ll A$ (i.e. low sampling intensity)

$$SE(\widehat{N}) = \frac{A}{a} \sqrt{\sum n_i}$$

remember: $n_s = \sum n_i$ is the total sample count

in our example: $SE = 100^2 / (10 \times 10^2) \sqrt{30} = 54.8$

2. If you are NOT resampling previously sampled locations:

$$SE(\widehat{N}) = \frac{A}{a} \sqrt{\sum n_i (1 - a_s/A)}$$

This is the **Finite Area Correction**. If $a = A$ - you sampled everything - SE goes to 0 as expected.

in our example: $SE = 54.5 \dots$ Almost no difference (because $a \ll A$).

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Some more complex formulae

from Fryxell book Chapter 12:

Table 13.3 Estimates and their standard errors for animals counted on transects, quadrats, or sections. The models are described in the text.

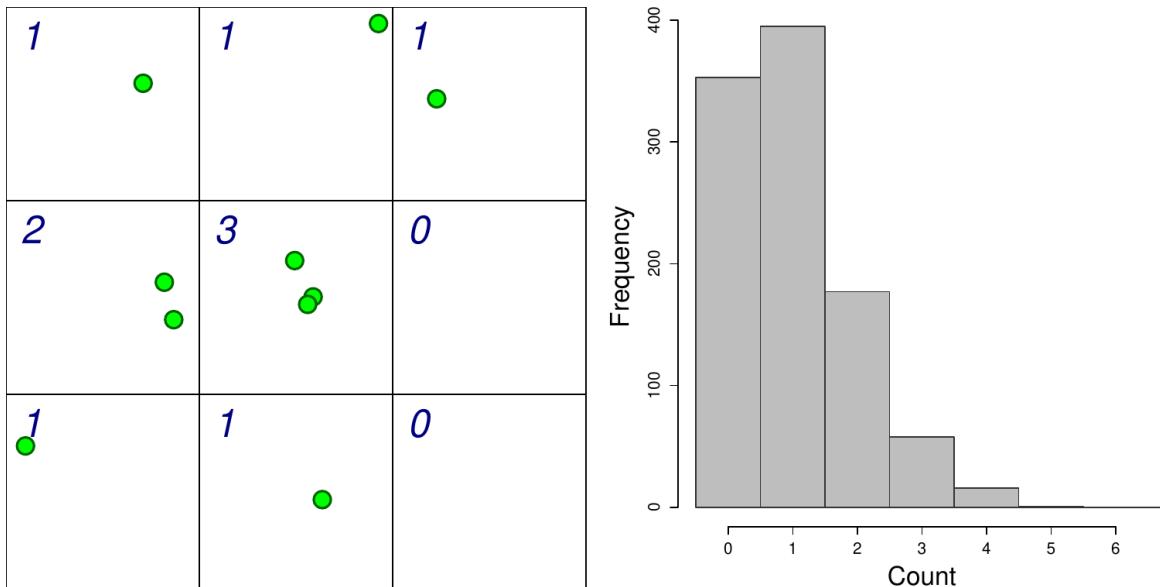
Model	Density	Numbers
<i>Simple</i>		
Estimate	$D = \sum y / \sum a$	$Y = A \times D$
Standard error of estimate (SWR)	$SE(D)_1 = 1/a \times \sqrt{[(\sum y^2 - (\sum y)^2/n) / (n(n-1))]}$	$SE(Y) = A \times SE(D)_1$
Standard error of estimate (SWOR)	$SE(D)_2 = SE(D)_1 \times \sqrt{[1 - (\sum a)/A]}$	$SE(Y) = A \times SE(D)_2$
<i>Ratio</i>		
Estimate	$D = \sum y / \sum a$	$Y = A \times D$
Standard error of estimate (SWR)	$SE(D)_3 = n / \sum a \times \sqrt{[(1/n(n-1))(\sum y^2 + D^2 \sum a^2 - 2D \sum ay)]}$	$SE(Y) = A \times SE(D)_3$
Standard error of estimate (SWOR)	$SE(D)_4 = SE(D)_3 \times \sqrt{[1 - (\sum a)/A]}$	$SE(Y) = A \times SE(D)_4$
<i>PPS</i>		
Estimate	$d = 1/n \times \sum (y/a)$	$Y = A \times d$
Standard error of estimate (SWR)	$SE(D) = \sqrt{[(\sum (y/a)^2 - (\sum (y/a))^2/n) / (n(n-1))]}$	$SE(Y) = A \times SE(d)$

SWR, sampling with replacement; SWOR, sampling without replacement. Notation is given in Section 13.5.1.

These are used when **sampling areas** are unequal, and account for differences when sampling **with replacement** or **without replacement**.

Poisson process

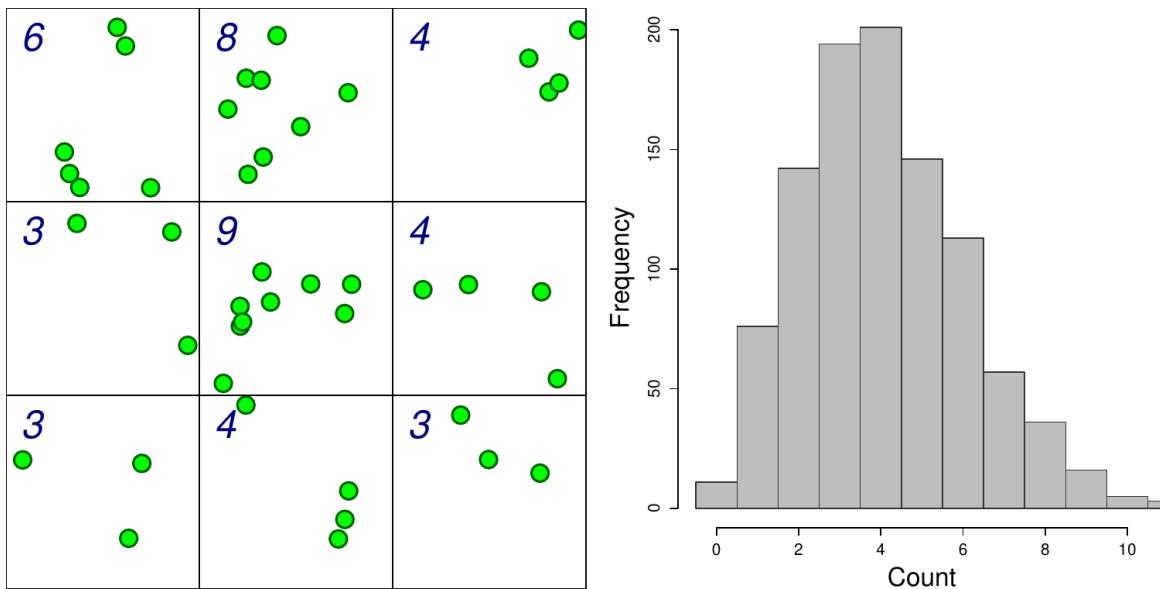
Models *counts*. If you have a perfectly random process with mean *density* (aka *intensity*) 1, you might have some 0 counts, you might have some higher counts. The *average* will be 1:



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Poisson process

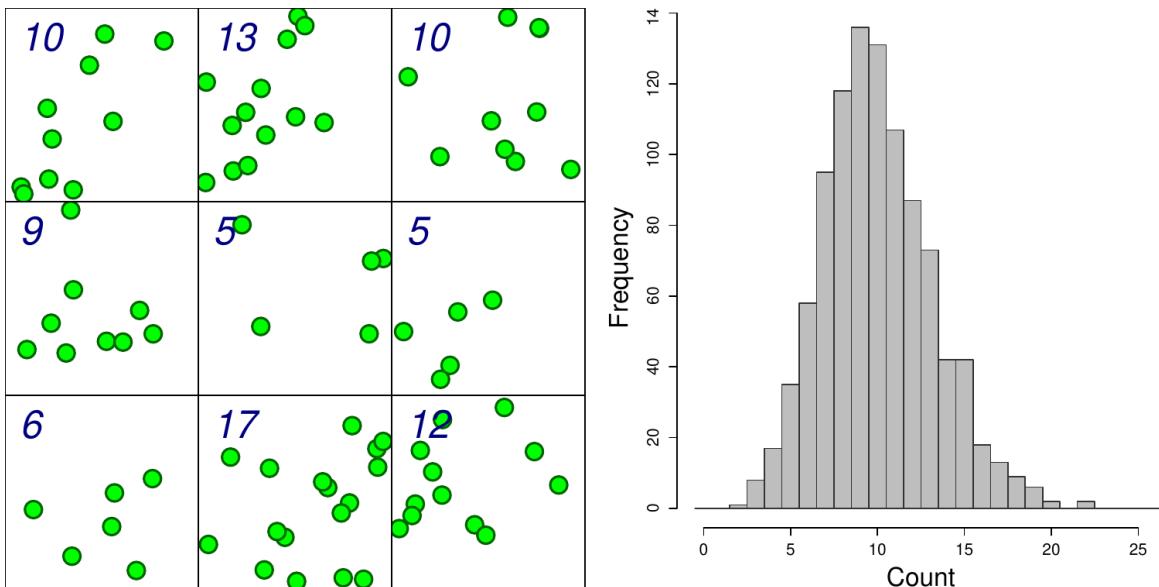
Here, the intensity is 4 ...



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Poisson process

... and 10. Note, the bigger the intensity, the more "bell-shaped" the curve.

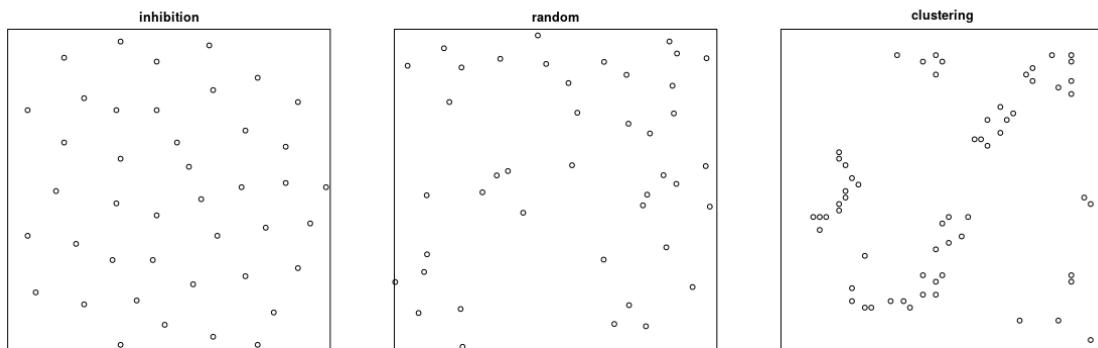


Here's the formula of the Poisson Distribution: $f(k; \lambda) = \Pr(X=k) = \frac{\lambda^k e^{-\lambda}}{k!}$

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Poisson distribution holds if process is truly random

... not **clustered** or **inhibited**

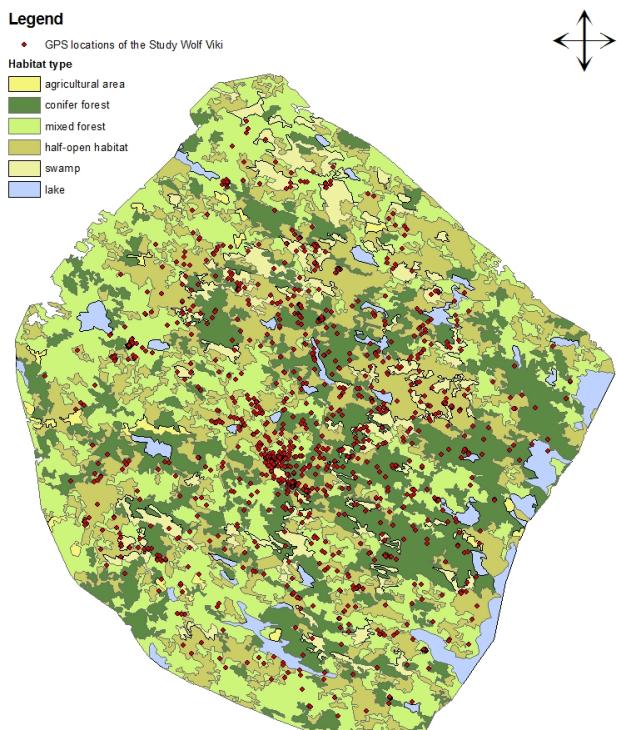


If you **sample** from these kinds of spatial distributions, your standard error might be smaller (*inhibited*) or larger (*clustering*). This is called *dispersion*.

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Also ... densities of animals can depend on habitat

Wolf habitat use

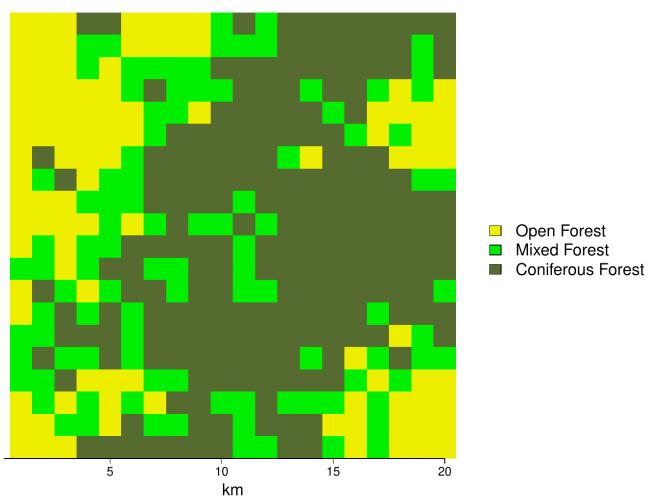


If you look closely:

- No locations in lakes
- Relatively few in bogs / cultivated areas.
- Quite a few in mixed and coniferous forest

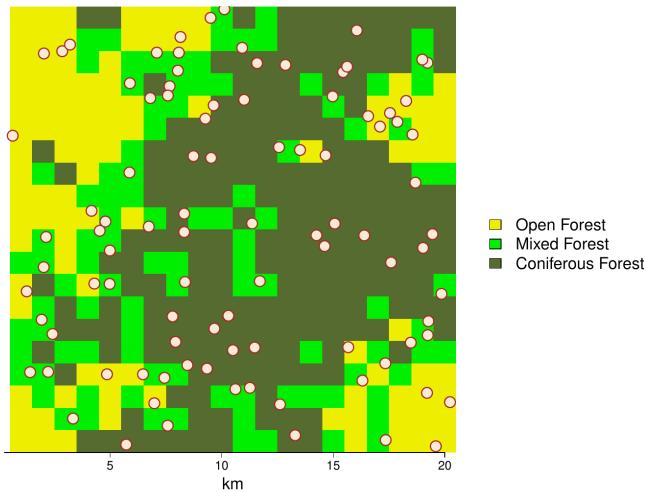
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Imagine a section of forest ...



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... with observations of moose



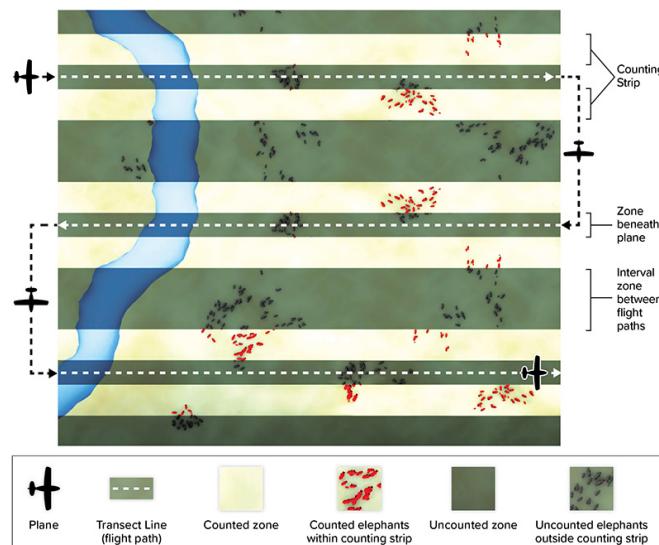
How can we tell what the moose prefers?

Habitat	Area	n	Density
open	100	21	0.21
mixed	100	43	0.43
dense	200	31	0.17
total	400	95	0.24

Knowing how densities differ as a function of **covariates** can be very important for generating estimates of abundances, increasing both **accuracy** and **precision**, and informing **survey design**.

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Sample frames need not be squares



Transects

Linear strip, usually from an aerial survey.

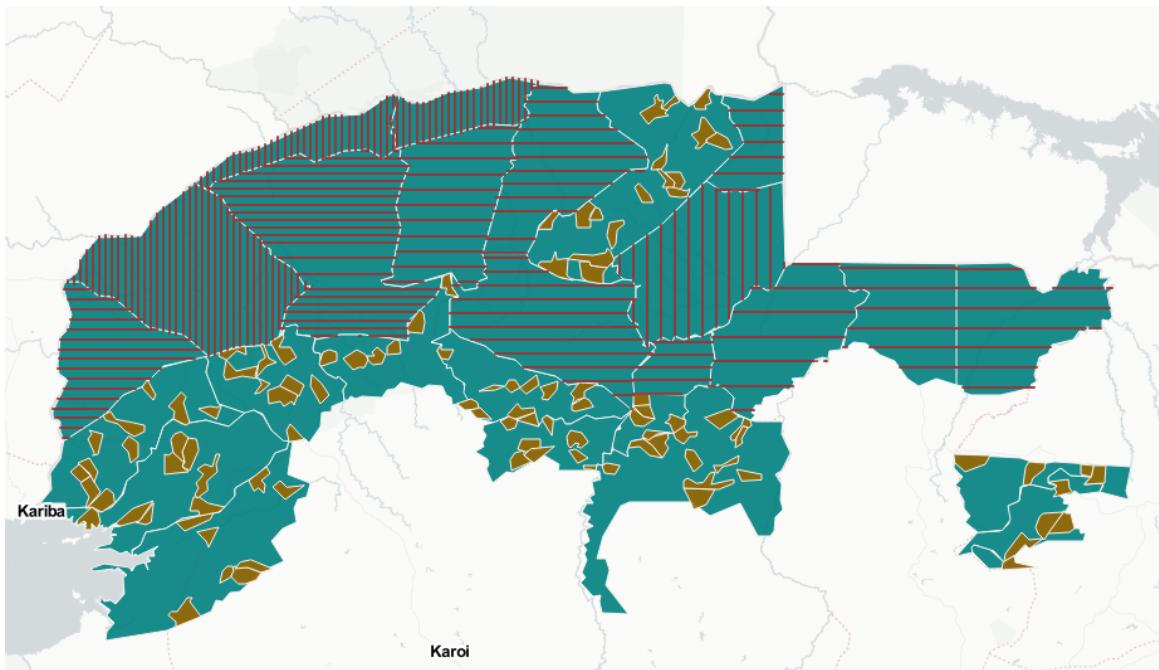
Efficient way to sample a lot of territory.

If "perfect detection", referred to as a **strip transect**.

Statistics - essentially - identical to quadrat sampling.

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Stratified sampling for more efficient estimation

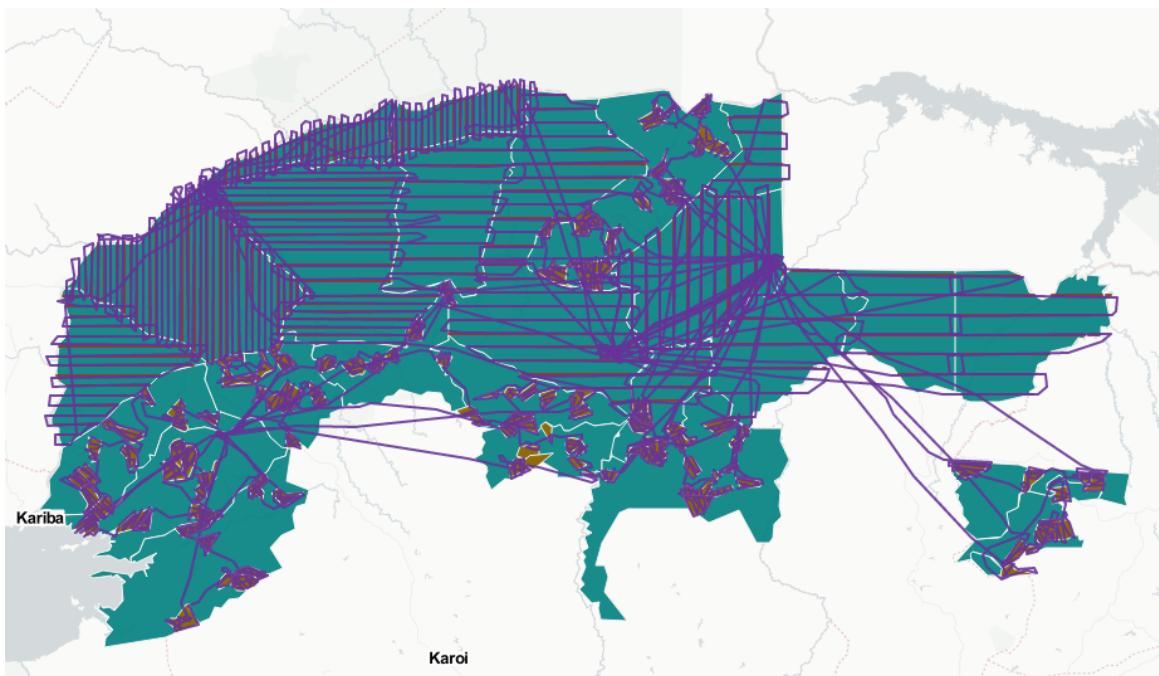


Sample more intensely in those habitats where animals are more likely to be found. Intensely survey **blocks** where detection is more difficult.

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<https://media.hhmi.org/bioInteractive/click/elephants/survey/survey-aerial-surveys-methods.html>

Stratified sampling for more efficient estimation

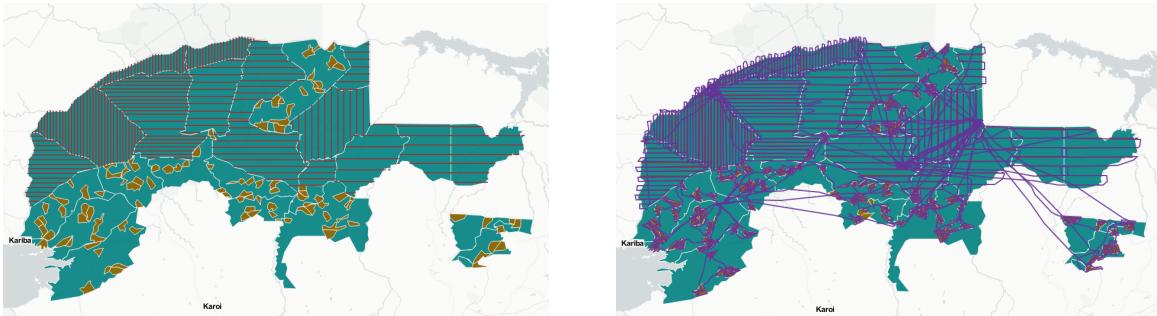


Actual elephant flight paths,

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<https://media.hhmi.org/bioInteractive/click/elephants/survey/survey-aerial-surveys-methods.html>

Stratified sampling

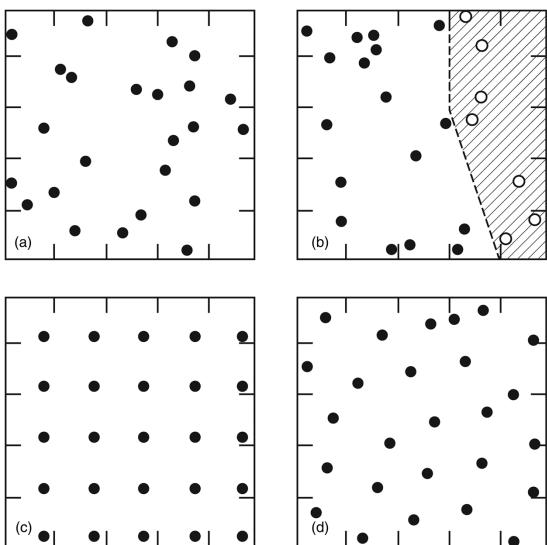


Stratification is used to optimize **effort** and **precision**. Aircraft cost thousands of dollars per hour!

(In all of these comprehensive surveys - *design* takes care of **accuracy**).

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Sampling strategies



- (a) simple random,
- (b) stratified random,
- (c) systematic,
- (d) pseudo-random (systematic unaligned).

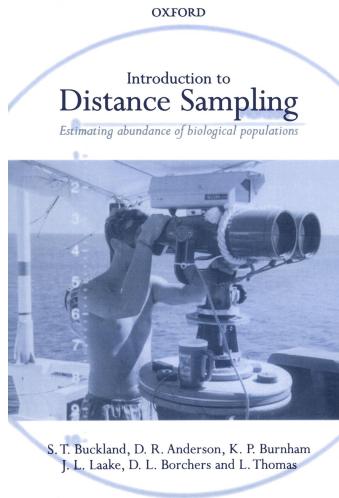
Each has advantages and disadvantages.

See also: *Adaptive Sampling*

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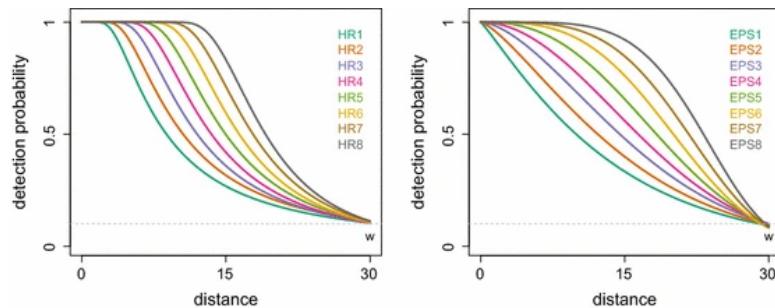
Detections usually get worse with distance!

Detection function	Form
Uniform	$1/w$
Half-normal	$\exp(-y^2/2\sigma^2)$
Hazard-rate	$1-\exp(-(y/\sigma)^b)$
Negative exponential	$\exp(-ay)$



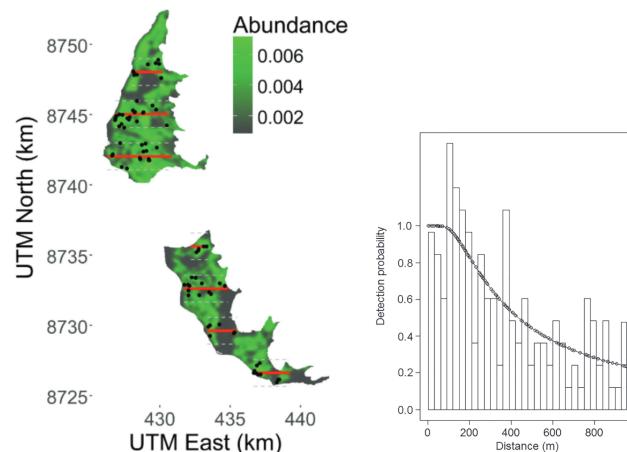
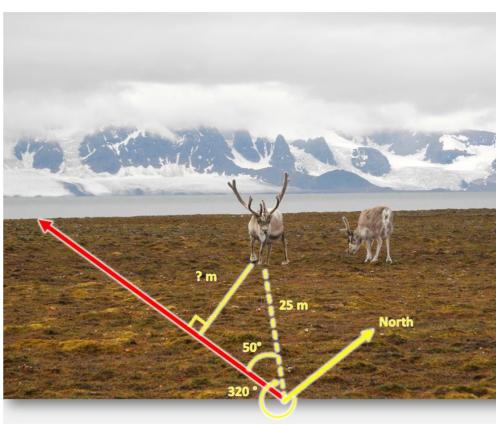
Distance Sampling

The statistics of accounting for visibility decreasing with distance



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Example reindeer in Svalbard



Ungulate population monitoring in an open tundra landscape:
distance sampling versus total counts

Wildlife Biology 2017; wlb.00299
doi: 10.2981/wlb.00299

Mathilde Le Moullec, Åshild Ønvik Pedersen, Nigel Gilles Yoccoz, Ronny Aanes, Jarle Tufto
and Brage Bremset Hansen

Estimated detection distance, compared to total count, incorporated vegetation modeling, computed standard errors, concluded that you can get a 15% C.V. for 1/2 the cost.

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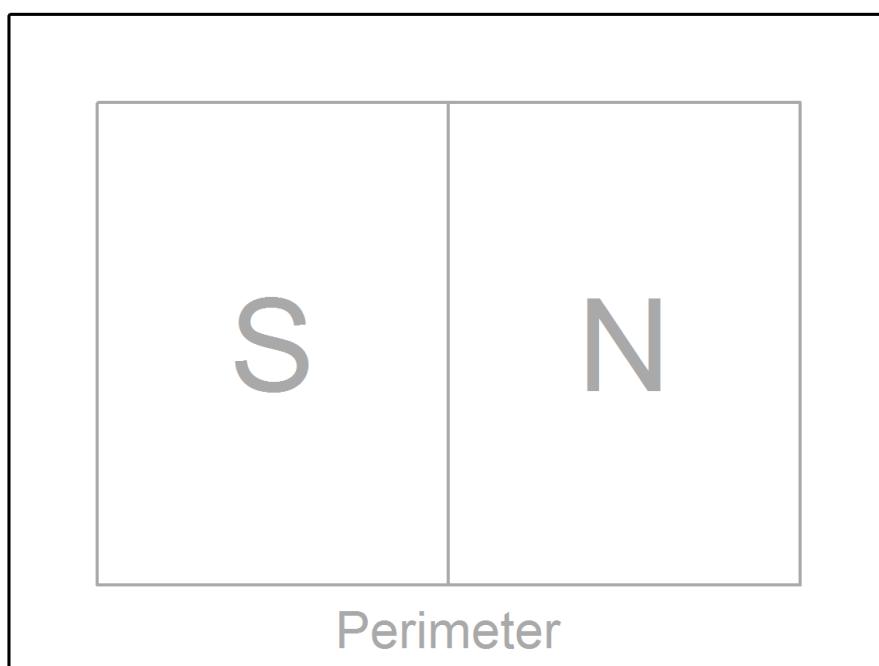
Example Ice-Seals

TRIBE LOBODONTINI



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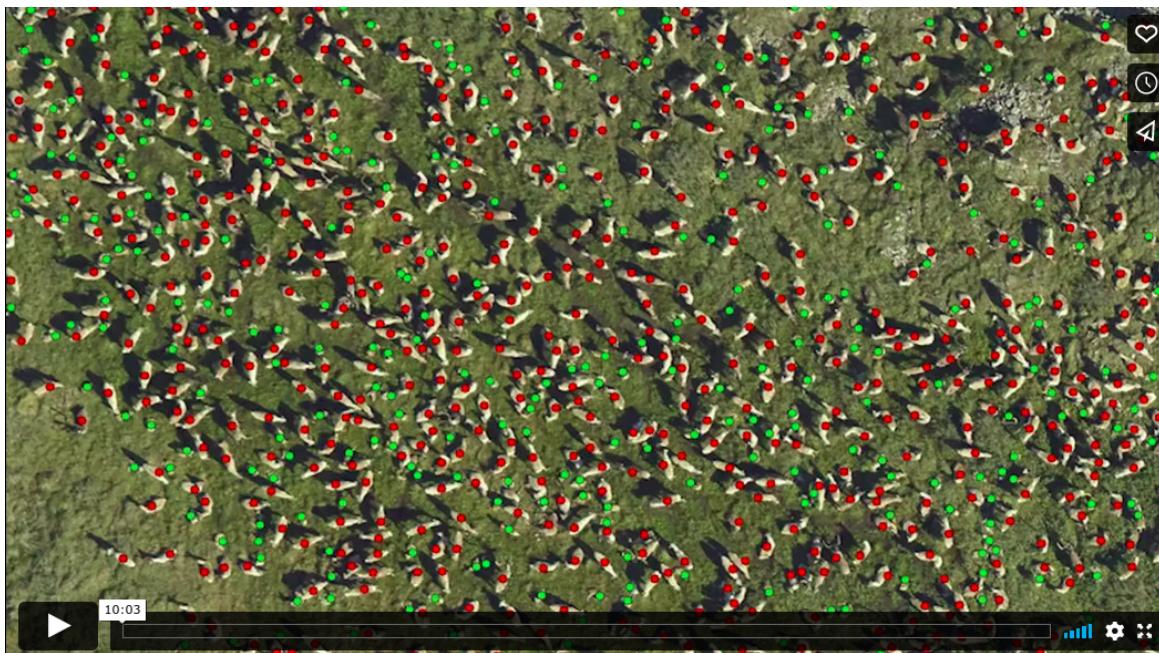
Example: Flag Counting at Baker



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Nice video on counting caribou

<https://vimeo.com/471257951>



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