

# Analysis of Stratified Surveys

Section 3.7 of Buckland et al. (2001)

Section 2.3 of Buckland et al. (2015)

# Stratification

- Why stratify?
- Stratification by:
  - Geographic area
  - Survey
  - Species / cluster size
- Decisions during analysis
- Alternatives to stratification

## Stratification is used to:

- reduce variance and improve precision
- and for producing estimates in regions of interest

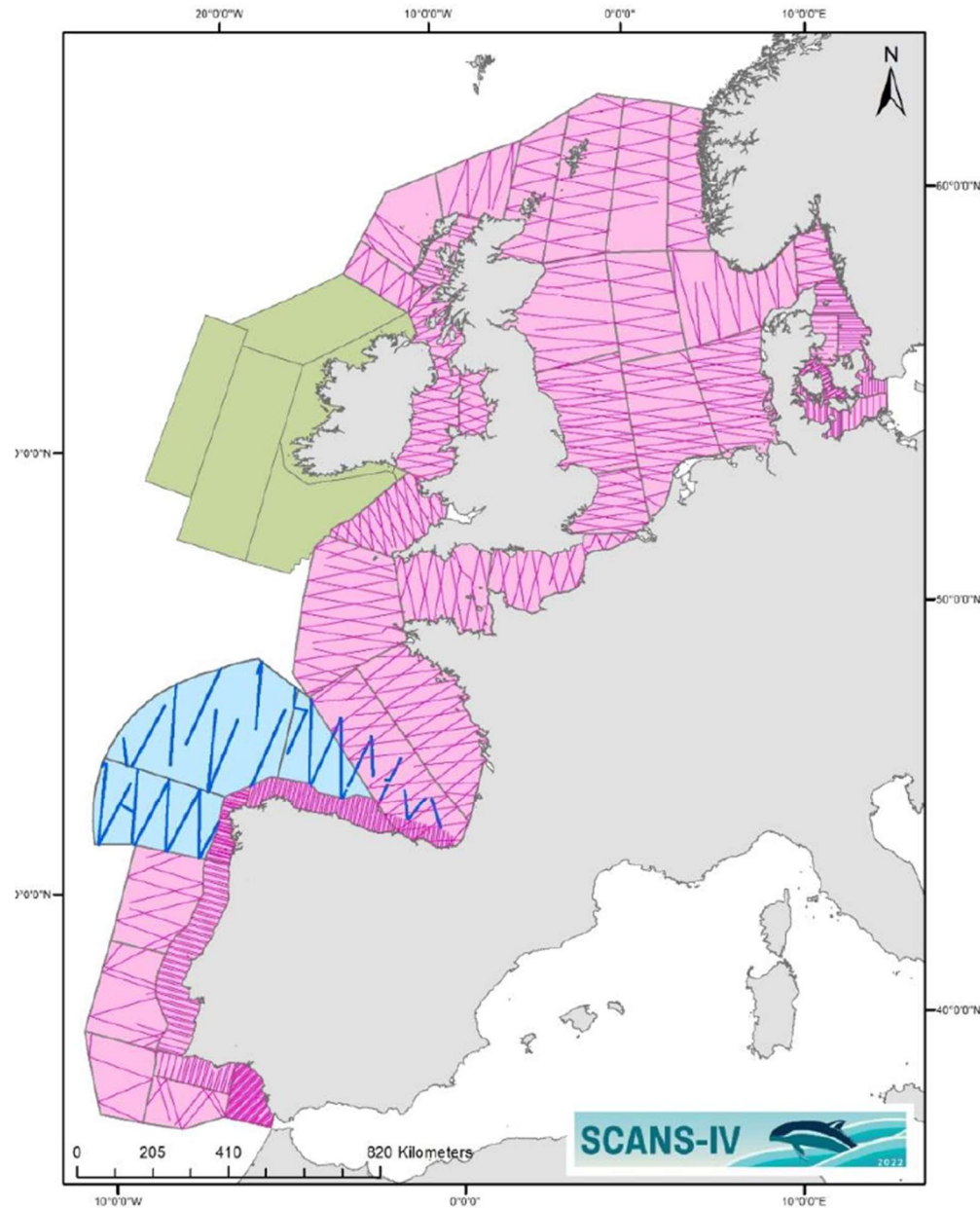
## Stratification criteria:

- AREA or GEOGRAPHIC REGION
  - the study region is partitioned into smaller regions
- SURVEY
  - used when different surveys cover the same geographic area
- POPULATION/SPECIES/CLUSTER SIZE
  - same geographic region containing different 'sub-stocks'

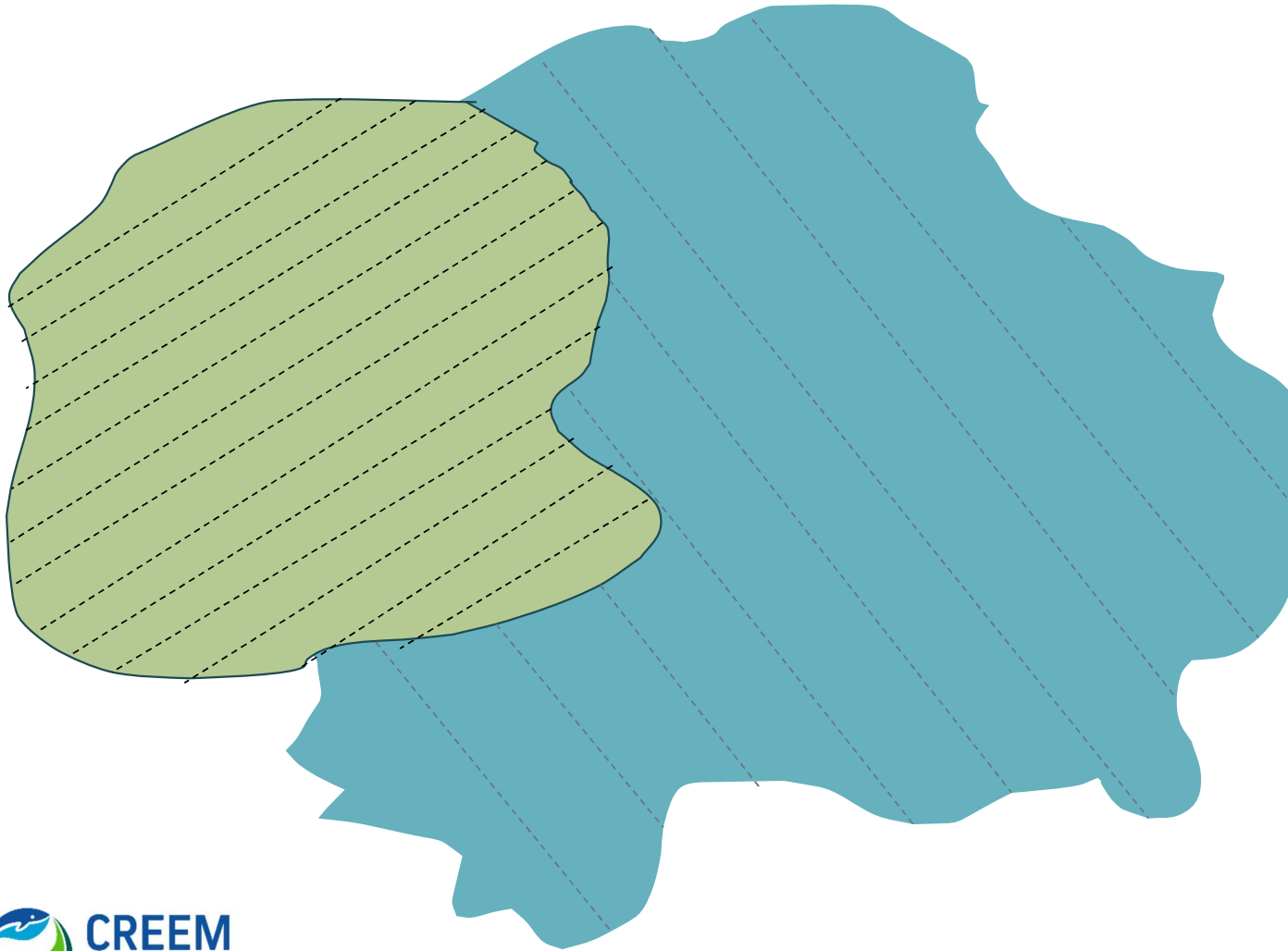
# Types of stratification

# Example geographic stratification: SCANS IV (2022)

Small Cetaceans in European Atlantic waters and the North Sea



# Geographic stratification



- Strata are geographic areas.
- Density estimates are required for each stratum and for the entire study area.

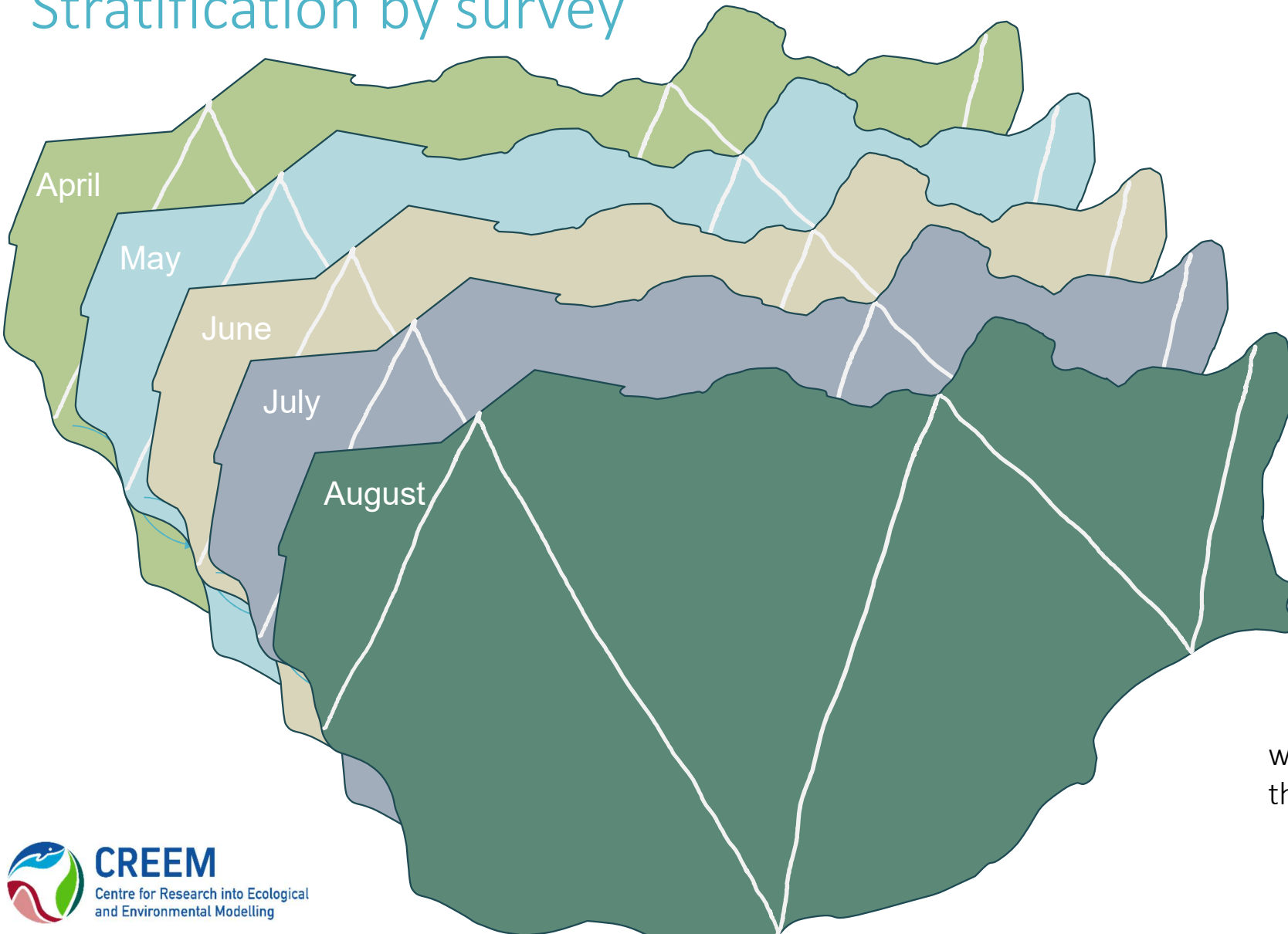
$$\hat{D} = \sum_{i=1}^m \frac{A_m}{A_{total}} \hat{D}_m$$

$$var(\hat{D}) = \sum_{i=1}^m \left( \frac{A_m}{A_{total}} \right)^2 var(\hat{D}_m)$$

where

- $\hat{D}_i$  is estimated density for the  $i^{\text{th}}$  stratum and
- $A_i$  is area of the  $i^{\text{th}}$  stratum

# Stratification by survey



- Replicate surveys have been conducted; e.g. week-long surveys conducted monthly or concurrently by different platforms.
- Interest lies in the average density across surveys and variability between surveys.

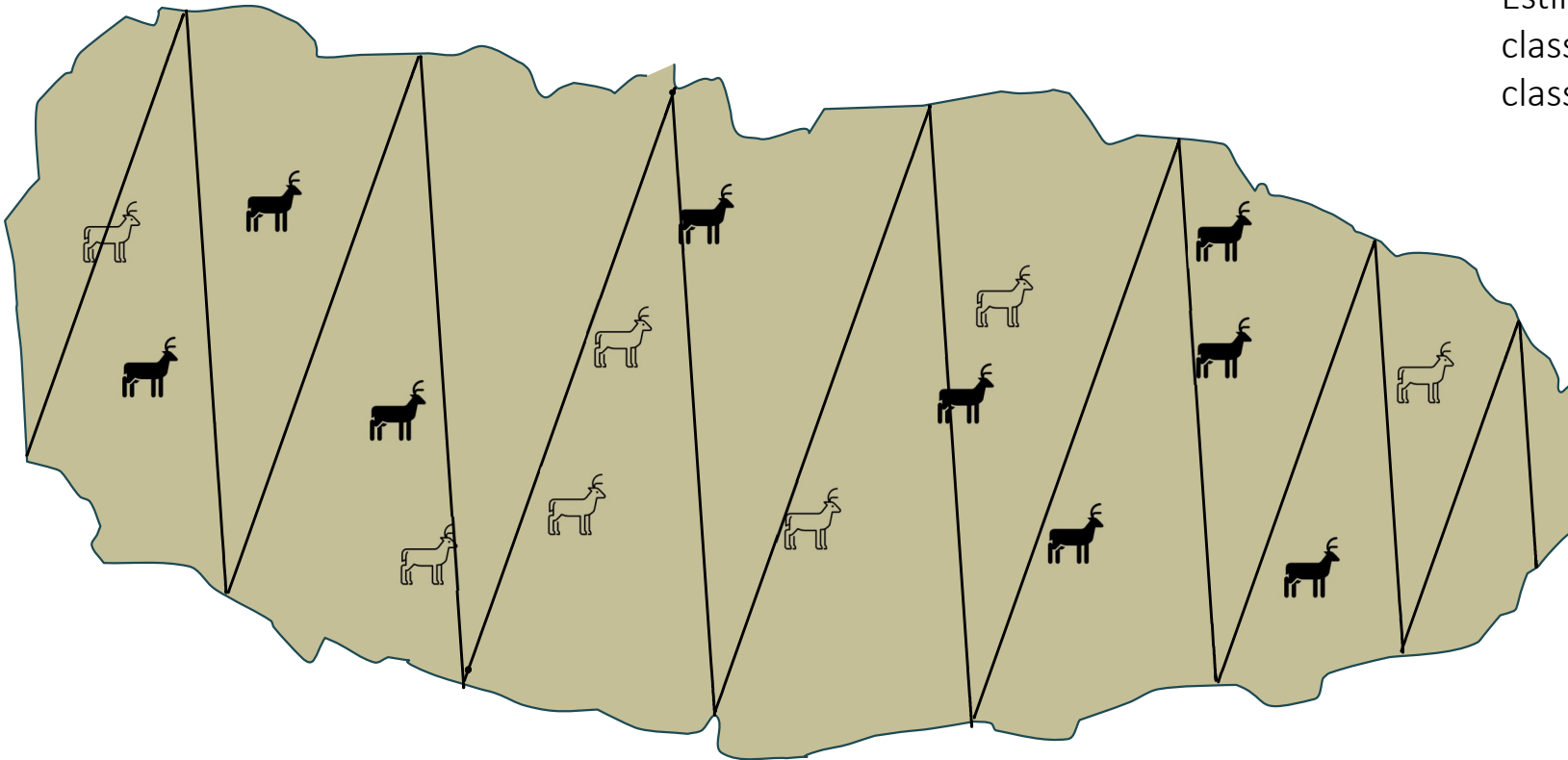
$$\hat{D} = \sum_{i=1}^M \frac{L_m}{\sum_{k=1}^M L_k} \hat{D}_m$$

$$\text{var}(\hat{D}) = \sum_{m=1}^M \frac{(\hat{D}_m - \bar{D})^2}{M - 1}$$

where  $L_i$  is effort associated with the  $i^{\text{th}}$  survey

## Post-stratification (stratification by object class)

- Objects are of different species or sexes.
- Estimates are desired for each object class as well as a total density across classes.

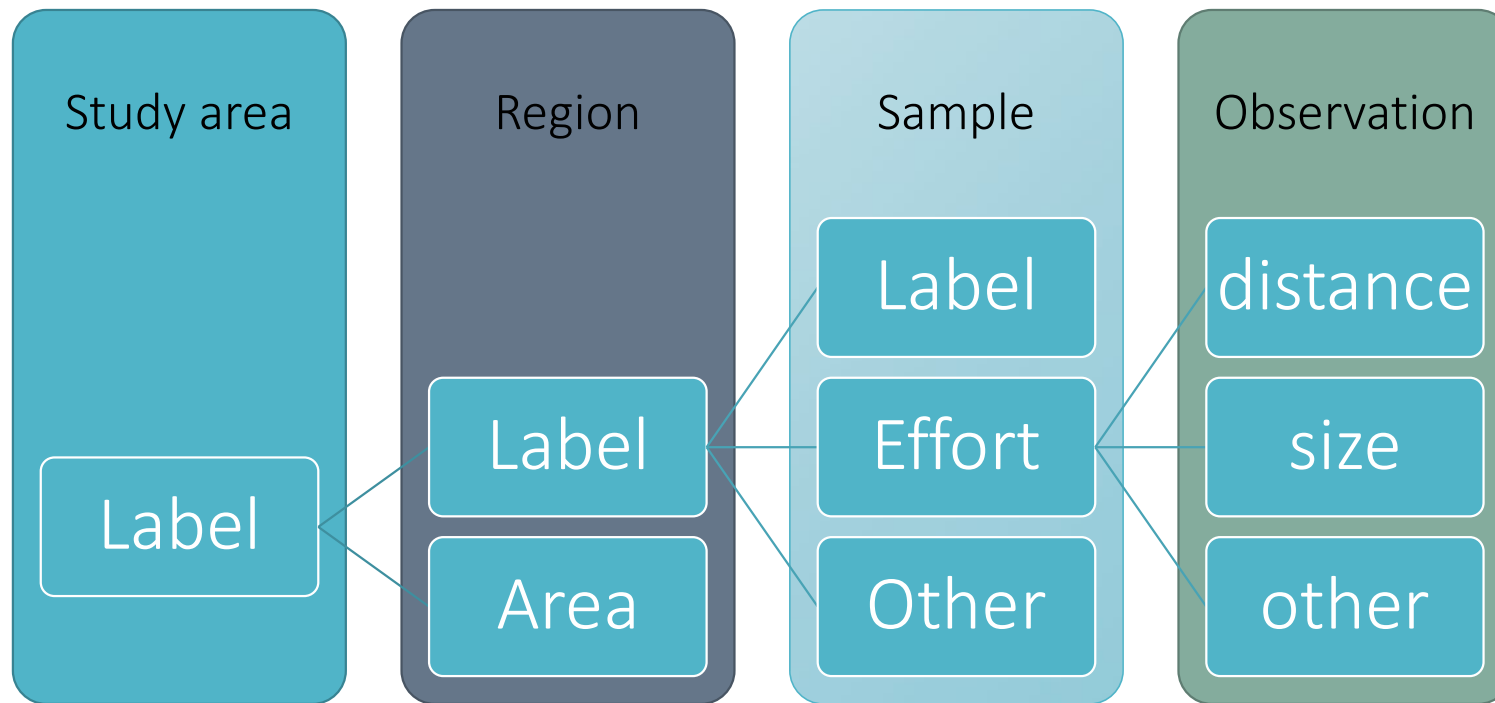


$$\hat{D} = \sum_{m=1}^M \hat{D}_m$$

$$\text{var}(\hat{D}) = \sum_{m=1}^M \text{var}(\hat{D}_m)$$



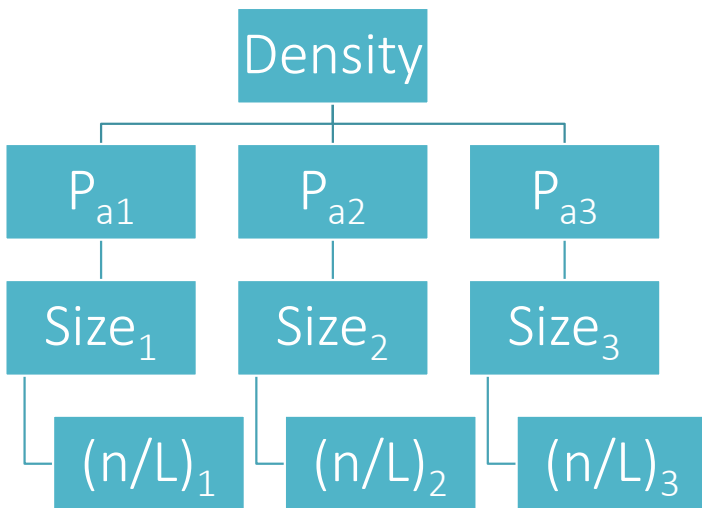
# Data organisation hierarchy



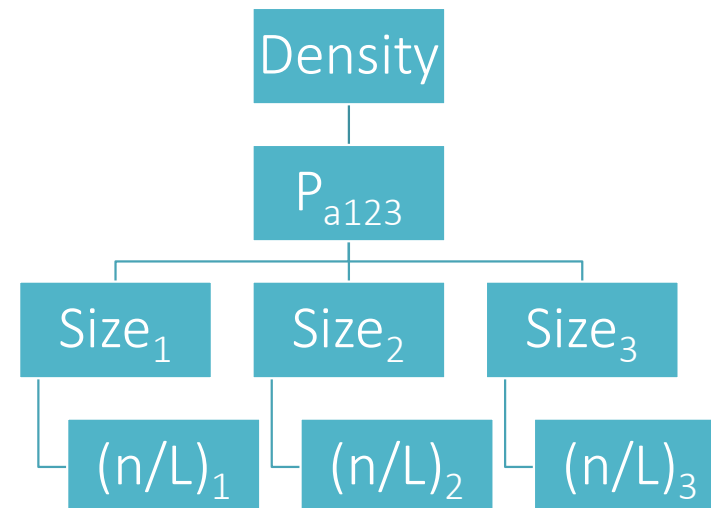
# Analysis decisions arising from stratification

## Example (3 strata):

Full geographic stratification



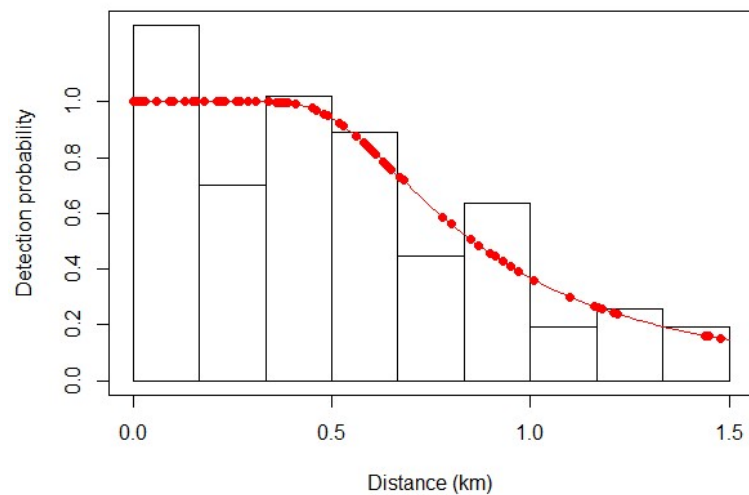
Pooled detection across strata



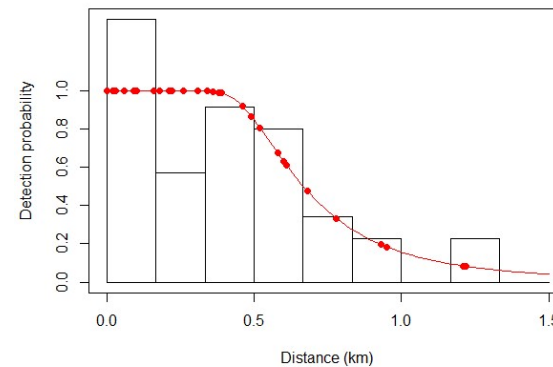
Select strata and fit detection function to each strata

# Pooled vs Stratified $P_a$

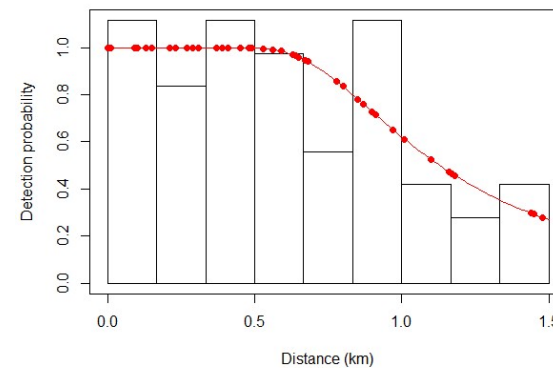
Pooled  $n=88$



Stratified  
Ideal habitat  $n=39$



Marginal habitat  $n=49$



# It is a Model Selection Problem

	Pooled	Stratum 1	Stratum 2	Stratum Sum
Log likelihood $\log_e(L)$	-180.490	-72.699	-104.676	-177.375
No. parameters ( $q$ )	2	2	2	4
AIC	364.980	149.398	213.352	362.75

Criterion for stratification of  $P_a$ :  
Fit separate  $P_a$  for each strata if

$$AIC_{pooled} > \sum_{strata} AIC_{stratum}$$

# Alternatives to stratification

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- Small sample sizes can lead to low precision in stratum-specific estimates
- An alternative approach to reducing bias due to heterogeneity is Multiple Covariates Distance Sampling (MCDS)
  - Covariates, other than distance, are incorporated into the scale parameter of the detection function
- MCDS can be used to fit the detection function at multiple levels e.g. stratum-specific density estimates can be obtained even with insufficient data to fit separate detection functions for each stratum