What is a density surface model?



Why model abundance spatially?

- Use non-designed surveys
- Use environmental information
- Maps

Back to Horvitz-Thompson estimation

Horvitz-Thompson-like estimators

• Rescale the (flat) density and extrapolate

$$\hat{N} = \frac{\text{study area}}{\text{covered area}} \sum_{i=1}^{n} \frac{s_i}{p_i^2}$$

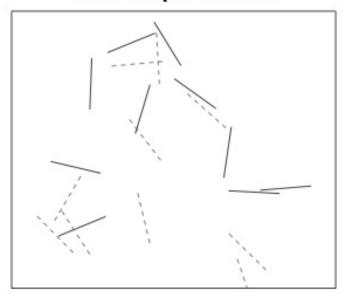
- s_i are group/cluster sizes
- p_i is the detection probability (from detection function)

Hidden in this formula is a simple assumption

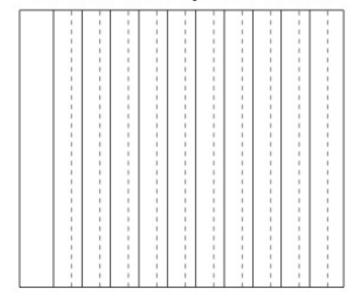
- Probability of sampling every point in the study area is equal
- Is this true? Sometimes.
- If (and only if) the design is randomised

Many faces of randomisation

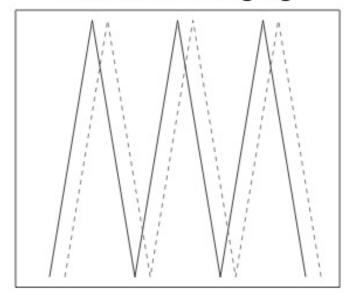
random placement



random offset parallel lines

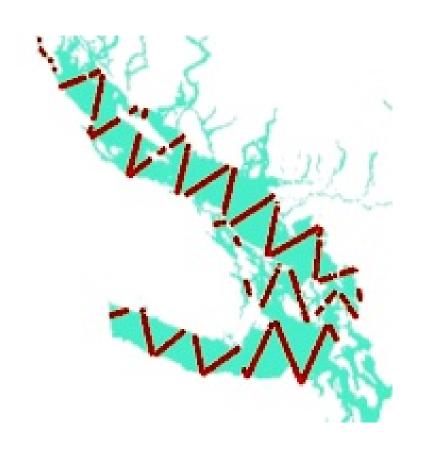


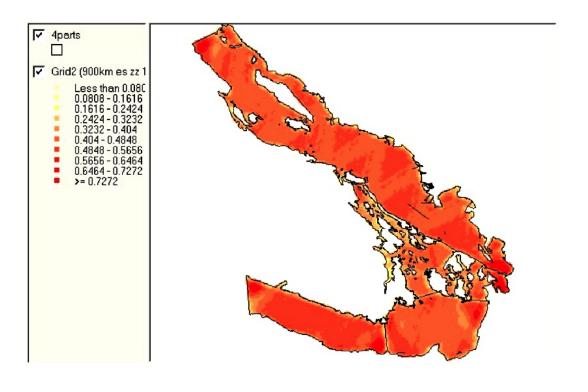
random offset zigzag



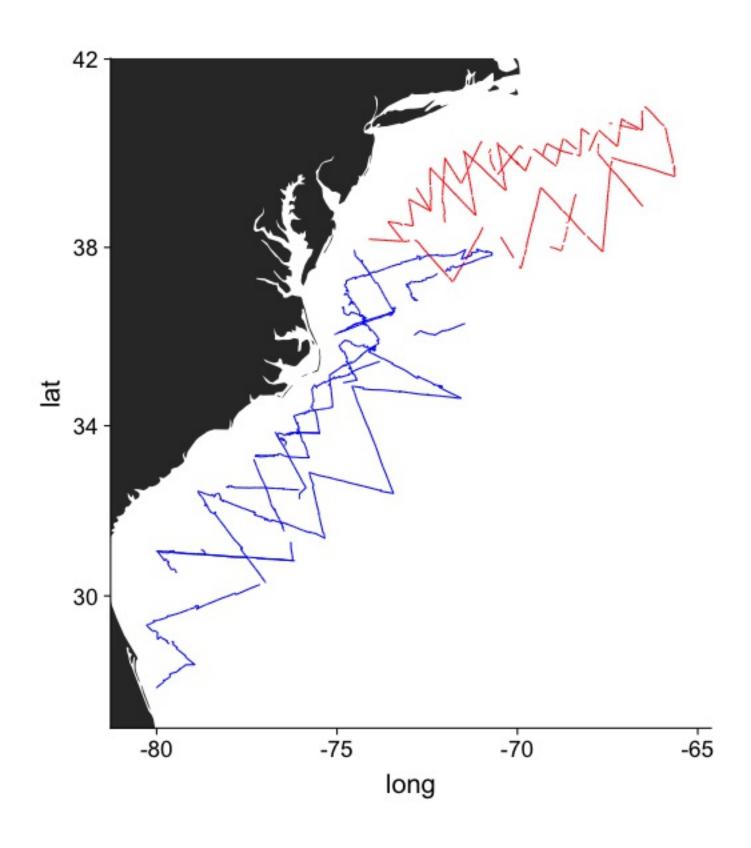
Randomisation & coverage probability

- H-T equation above assumes even coverage
 - (or you can estimate)

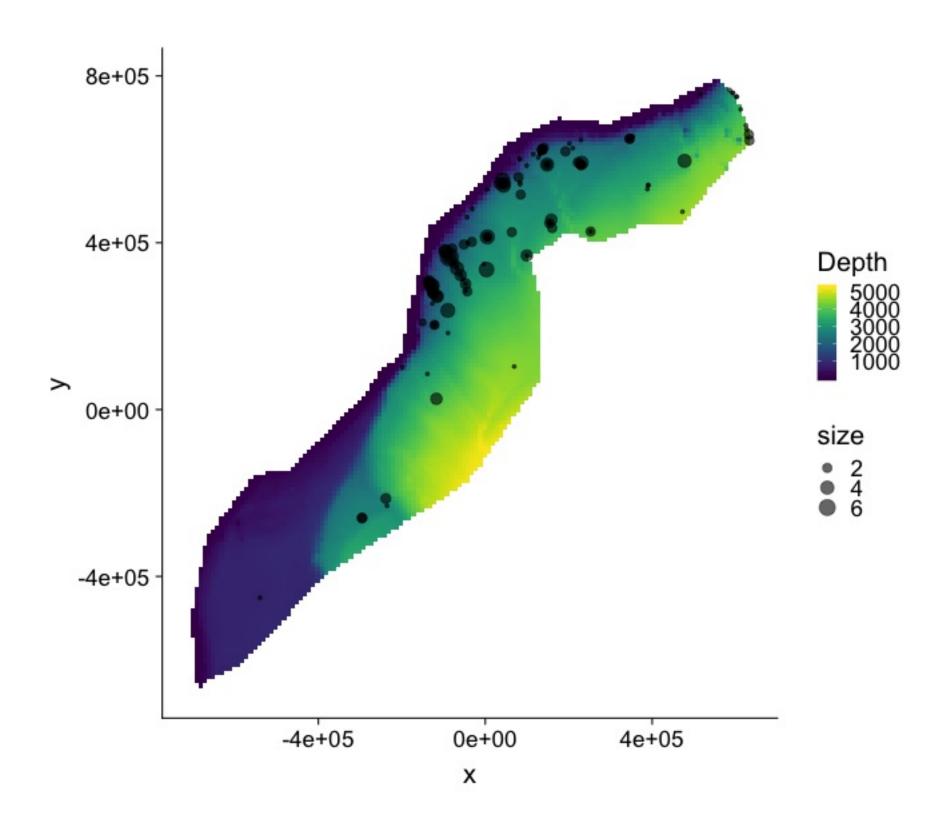




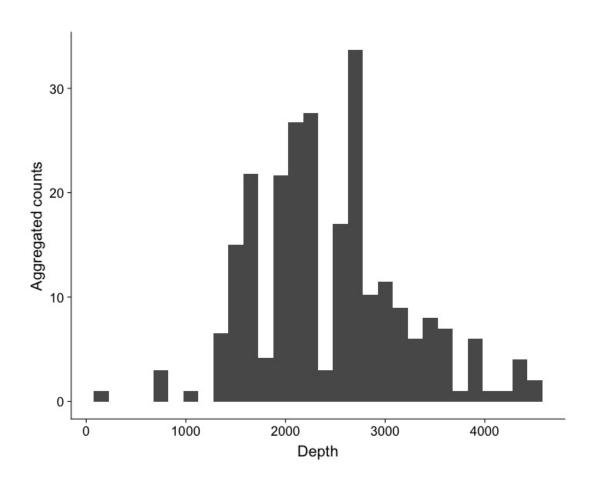
Extra information



Extra information - depth

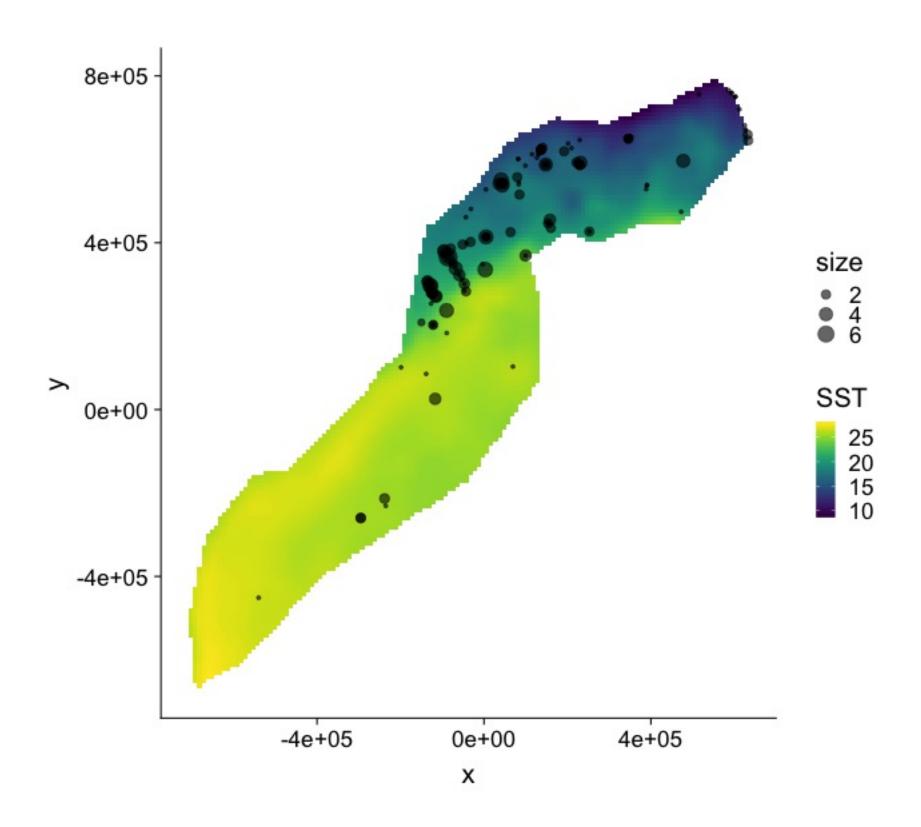


Extra information - depth

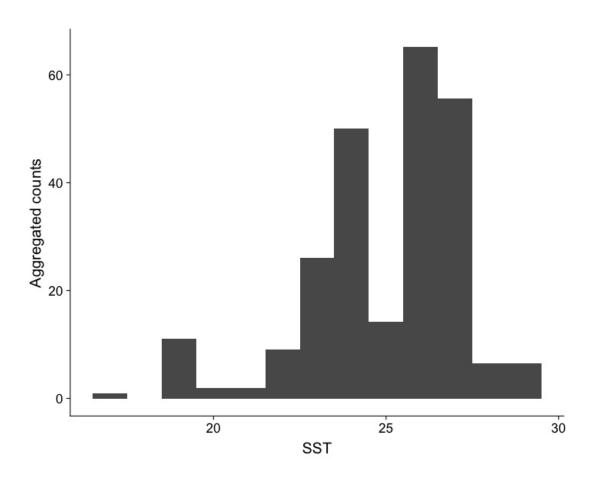


• NB this only shows segments where counts > 0

Extra information - SST



Extra information - SST



(only segments where counts > 0)

You should model that

Modelling outputs

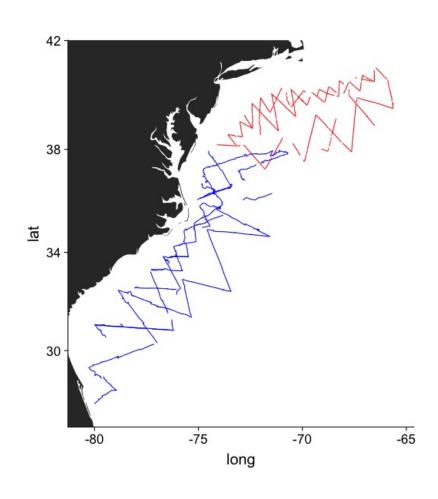
- Abundance and uncertainty
 - Arbitrary areas
 - Numeric values
 - Maps
 - Extrapolation (with caution!)
- Covariate effects
 - count/sample as function of covars

Modelling requirements

- Include detectability
- Account for effort
- Flexible/interpretable effects
- Predictions over an arbitrary area

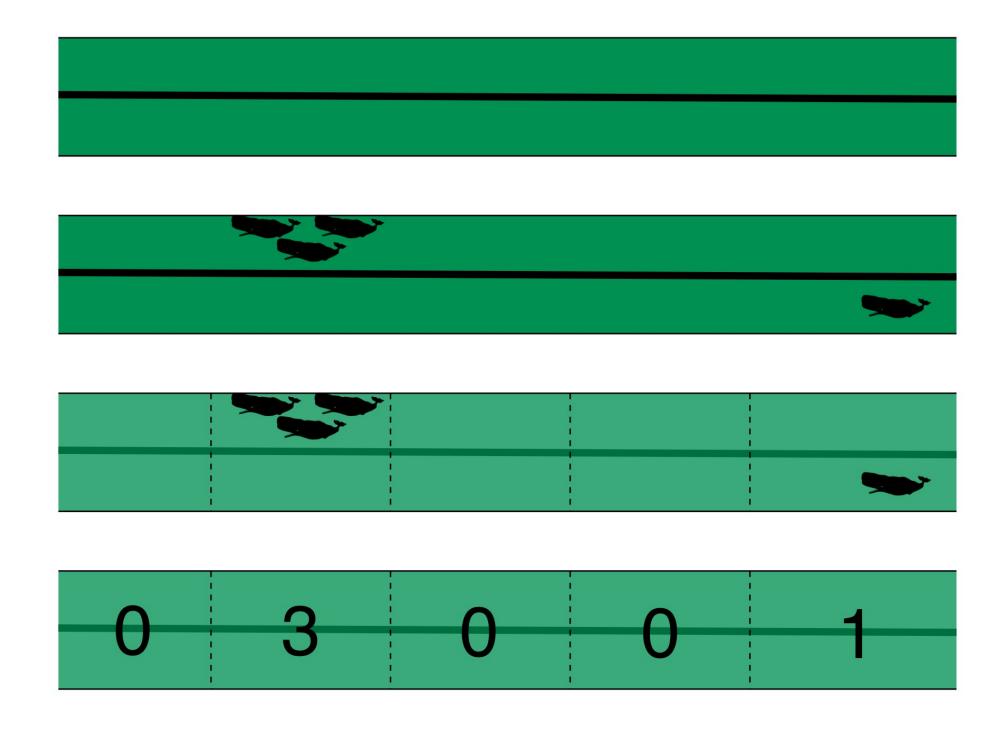
Accounting for effort

Effort



- Have transects
- Variation in counts and covars along them
- Want a sample unit w/ minimal variation
- "Segments": chunks of effort

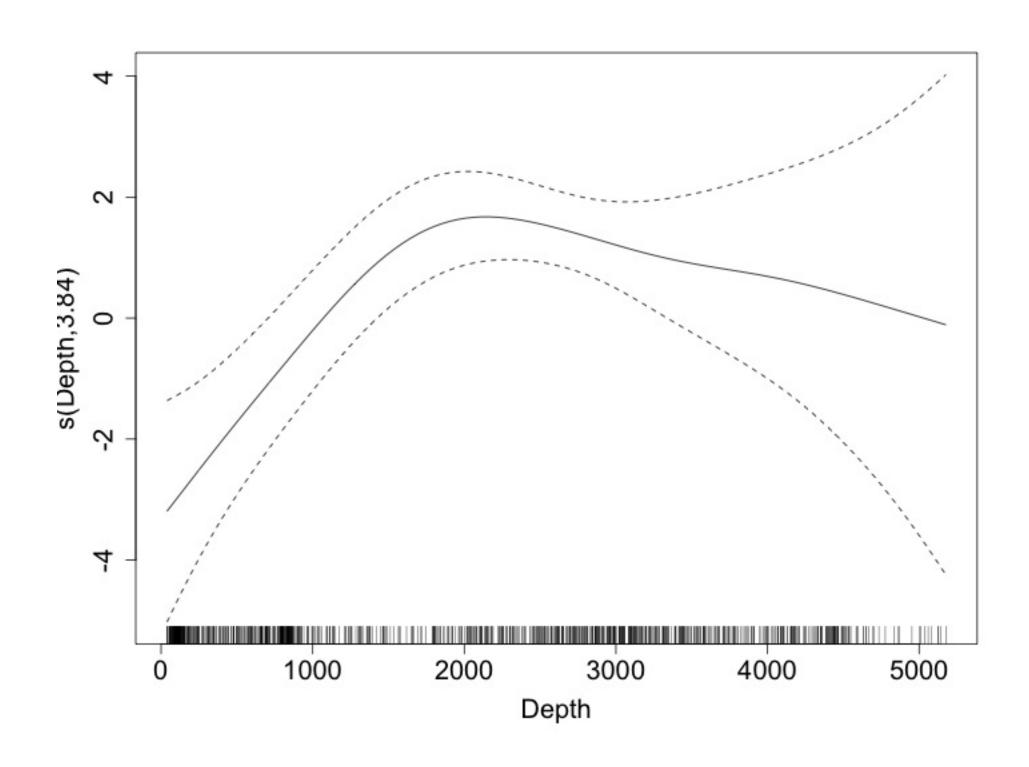
Chopping up transects



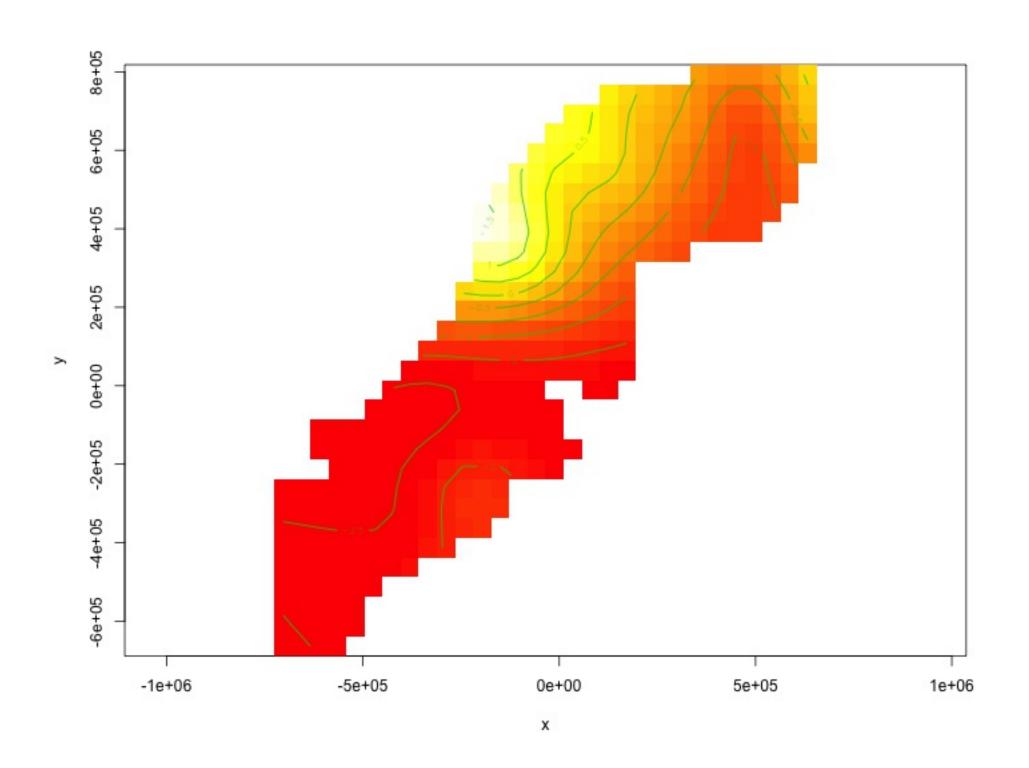
Physeter catodon by Noah Schlottman

Flexible, interpretable effects

Smooth response

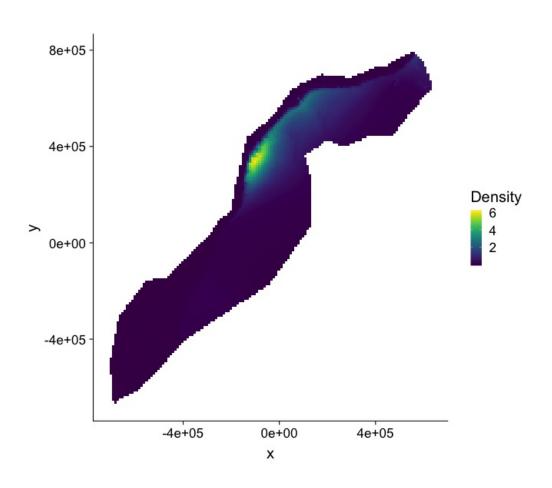


Explicit spatial effects



Predictions

Predictions over an arbitrary area



- Don't want to be restricted to predict on segments
- Predict within survey area
- Extrapolate outside (with caution)
- Working on a grid of cells

Detection information

Including detection information

- Two options:
 - adjust areas to account for effective effort
 - use **Horvitz-Thompson estimates** as response

Effective effort

- ullet Area of each segment, A_j
 - \circ use $A_j p_j^{\hat{}}$
- think effective strip width ($\hat{\mu} = wp$)
- Response is counts per segment
- "Adjusting for effort"
- "Count model"

Estimated abundance

- Estimate H-T abundance per segment
- Effort is area of each segment
- "Estimated abundance" per segment

$$\hat{n_j} = \sum_{i} \frac{s_i}{\hat{p_i}}$$

(where the i observations are in segment j)

Detectability and covariates

- 2 covariate "levels" in detection function
 - "Observer"/"observation" -- change within segment
 - "Segment" -- change between segments
- "Count model" only lets us use segment-level covariates
- "Estimated abundance" lets us use either

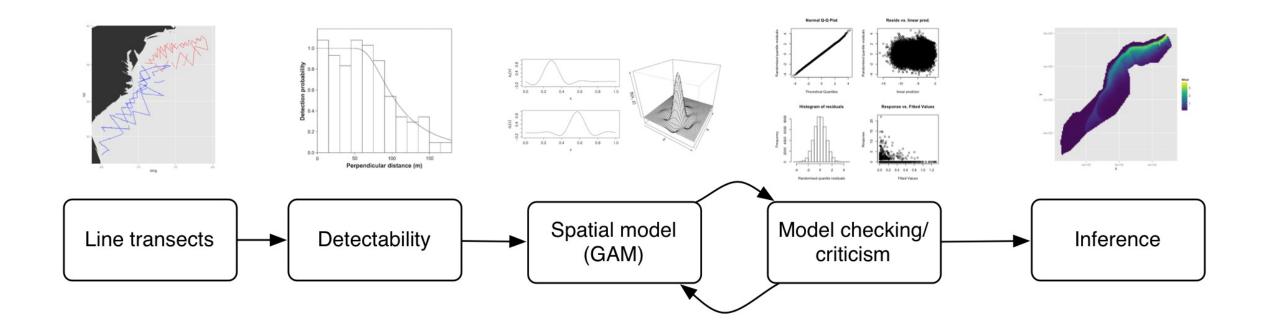
When to use each approach?

- Generally "nicer" to adjust effort
- Keep response (counts) close to what was observed
- Unless you want observation-level covariates

Availability, perception bias and more

- p is not always simple!
- Availability & perception bias somehow enter
- We can make explicit models for this
- More later in the course

DSM flow diagram



Spatial models

Abundance as a function of covariates

- Two approaches to model abundance
- Explicit spatial models
 - When: good coverage, fixed area
- "Habitat" models (no explicit spatial terms)
 - When: poorer coverage, extrapolation
- We'll cover both approaches here

Data requirements

What do we need?

- Need to "link" data
 - Distance data/detection function
 - Segment data
 - Observation data (segments detections)

Example of spatial data in QGIS

Recap

- Model counts or estimated abundance
- The effort is accounted for differently
- Flexible models are good
- Incorporate detectability
- 2 tables + detection function needed