Multiple covariate distance sampling (MCDS) - Complications

• Aim:

- Discuss some complications that arise with MCDS analyses
- Give some analysis guidelines





Complications 1. Clustered populations

There are two approaches to estimating number of individuals when objects are in clusters:

(1)
$$\hat{N} = \sum_{i=1}^{n} \frac{1}{\Pr[group\ included]} \hat{E}[s]$$

$$= \frac{A}{2L} \hat{E}(s) \sum_{i=1}^{n} \hat{f}(0 \mid z_i)$$
(2)
$$\hat{N} = \sum_{i=1}^{n} \frac{group\ size}{\Pr[group\ included]}$$

$$= \frac{A}{2L} \sum_{i=1}^{n} s_i \hat{f}(0 \mid z_i)$$

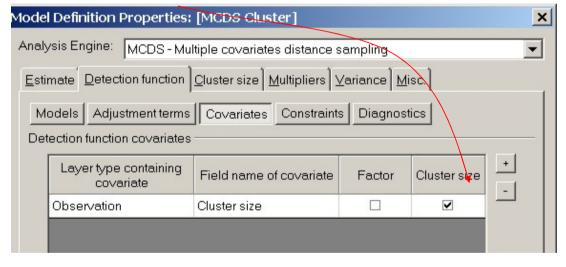
When cluster size is not a covariate, we use (1); when it is a covariate, we use (2)





Clustered populations (contd.)

To tell Distance that a covariate represents cluster size, tick the box:



When cluster size is a covariate:

- Distance does not estimate variance using analytic methods: the bootstrap must be used (Reflected in the Variance tab)
- •There is no need for size bias regression methods (Cluster size tab changes)
- No stratification allowed (Estimate tab)





Complications 2. Adjustment terms

With adjustments:
$$g(x,\mathbf{z}) = k(x,\mathbf{z}) \left[1 + \sum_{j=1}^{m} a_{j} p_{j}(x_{s})\right] / c$$

Adjustment terms use *scaled* distances, x_s

- cosine adjustment of order 2: $\cos(2\pi x_s)$
- simple polynomial of order 4: x_s^4

Why scale?

- Avoid numerical problems
- Limits cosine adjustment to a small number of 'wiggles'

How to scale?





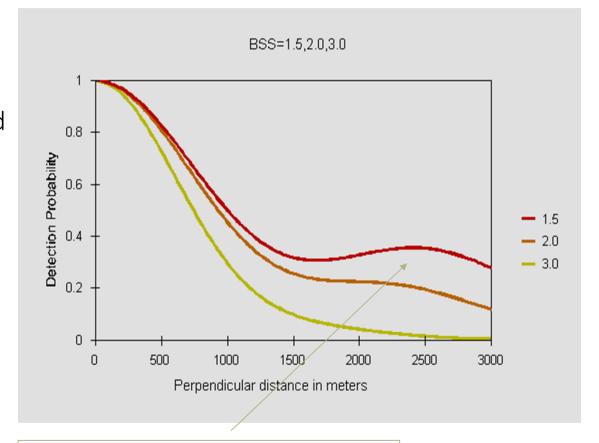
Adjustment terms (contd.)

Scenario 1: Scale distances by w, the right truncation distance $x_s = x/w$

Then covariates affect the scale of the key function, but adjustment terms are unaffected by covariates, so the overall shape varies with covariate value:

e.g., half-normal with 1 cosine adjustment of order 2

$$g(x \mid \mathbf{z}) \propto \exp\left(\frac{-x^2}{2\sigma(\mathbf{z})^2}\right) \left[1 + a_2 \cos\left(\frac{2\pi x}{w}\right)\right]$$









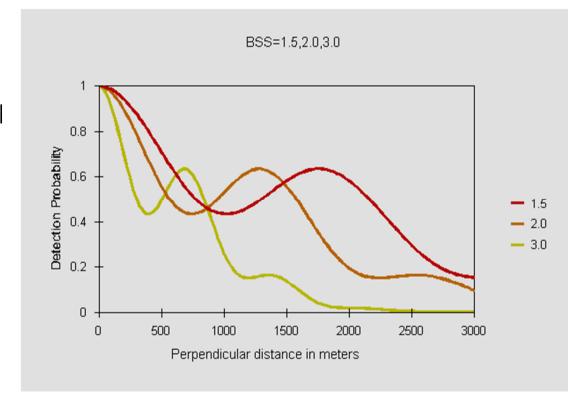
Adjustment terms (contd.)

Scenario 2: Scale distances by $\sigma(\mathbf{z})$, the estimated scale parameter $x_s = x/\sigma(\mathbf{z})$

Then covariates affect the scale of the key function, and the scale of the adjustment terms, so only the scale and not the shape of the overall function is affected:

e.g., half-normal with 1 cosine adjustment of order 2

$$g(x \mid z) \propto \exp\left(\frac{-x^2}{2\sigma(z)^2}\right) \left[1 + a_2 \cos\left(\frac{2\pi x}{\sigma(z)}\right)\right]$$

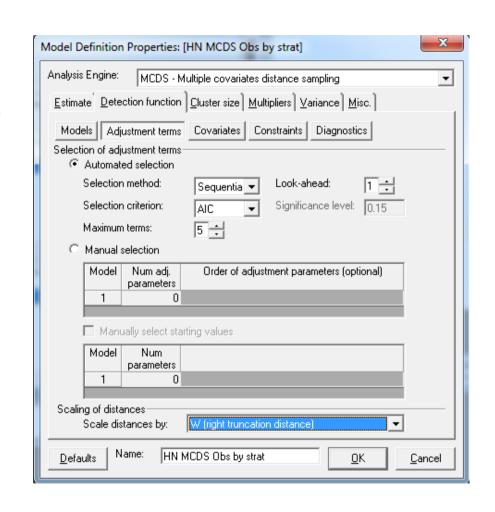






Adjustment terms (contd.)

- •The previous was an extreme example, to illustrate the difference between scaling factors.
- •Generally:
- start with no adjustment terms and introduce covariate terms one by one
- check the fit with adjustments looks reasonable
- consider whether to scale by w or σ
- you may need fewer adjustment terms with MCDS than CDS analyses

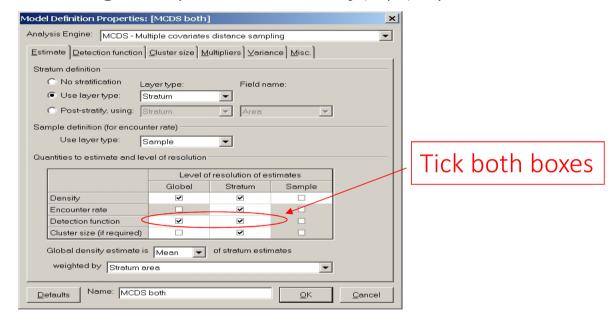






Complications 3. Stratification

If we want stratum-level estimates of density/abundance we can fit the detection function with covariates globally, and estimate $f(0|\mathbf{z})$ by stratum:



• Note: Global variance estimate for density/abundance must be calculated via the bootstrap





MCDS analysis guidelines

Choose covariates that are:

- independent of distance
- not strongly correlated with each other

Specifying the model:

- factor covariates generally harder to fit
- avoid or limit automatic selection of adjustment terms
- if using adjustments, consider whether to scale by w or σ
- check convergence and monotonicity
- add only one covariate at a time
- where necessary, use starting values and bounds for parameters
- consider reducing the truncation distance, w, if more than 5% of the $P_a(z_i)$ are <0.2, or if any are less than 0.1



