Lecture 5: Predictions and variance



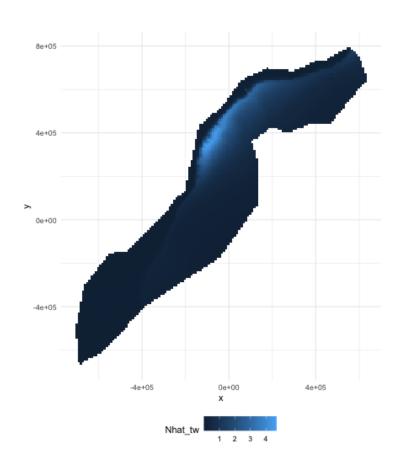
So far...

- Build, check & select detection models
- Build, check & select *spatial* models

What about predictions?

Let's talk about maps

What does a map mean?



- Grids!
- Cells are abundance estimate
- "snapshot"
- Sum cells to get abundance
- Sum a subset?

Going back to the formula

Count model (*j* observations):

$$n_j = A_j \hat{p}_j \exp[\beta_0 + s(y_j) + s(Depth_j)] + \epsilon_j$$

Predictions (index r):

$$\hat{n}_r = A_r \exp[\hat{\beta}_0 + \hat{s}(y_r) + \hat{s}(Depth_r)]$$

Need to "fill-in" values for A_r , y_r and Depth_r.

Predicting

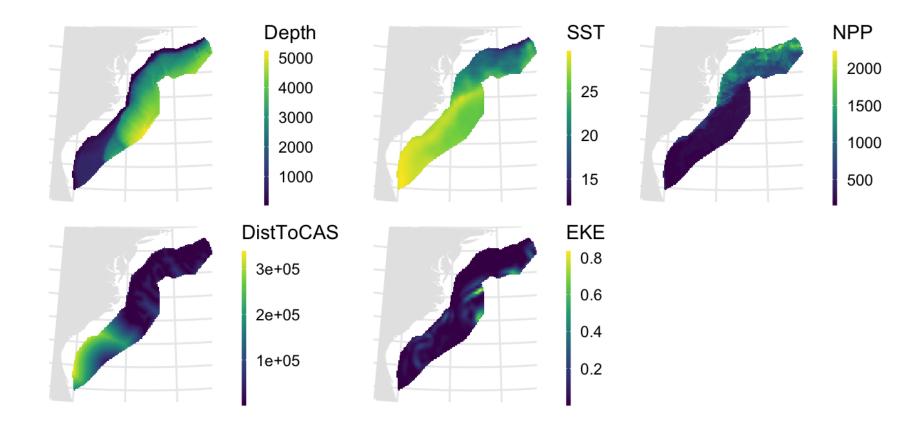
- With these values can use predict in R
- predict(model, newdata=data, off.set=off.set)

- off.set gives the area of the grid cells
- more info in ?predict.dsm

Prediction data

```
##
                       Depth SST NPP DistToCAS
  126 547984.6 788254 153.59825 12.04609 1462.521 11788.974
  127 557984.6 788254 552.31067 12.81379 1465.410
                                            5697,248
  258 527984.6 778254 96.81992 12.90251 1429.432 13722.626
  259 537984.6 778254 138.23763 13.21393 1424.862 9720.671
  260 547984.6 778254 505.14386 13.75655 1379.351 8018.690
  261 557984.6 778254 1317.59521 14.42525 1348.544
                                            3775,462
             EKE off.set
##
                           long
                                    lat
                  1e+08 -66.52252 40.94697
## 126 0.0008329031
## 127 0.0009806611 1e+08 -66.40464 40.94121
                  1e+08 -66.76551 40.86781
## 258 0.0011575423
## 260 0.0026881567 le+08 -66.52996 40.85662
```

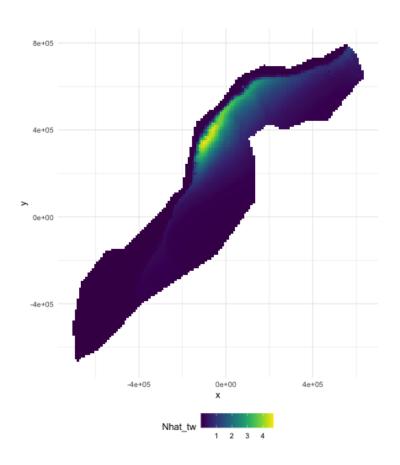
Predictors



Making a prediction

- Add another column to the prediction data
- Plotting then easier (in R)

Maps of predictions



Total abundance

Each cell has an abundance, sum to get total

```
sum(predgrid$Nhat_tw)
```

```
## [1] 2491.863
```

Subsetting

R subsetting lets you calculate "interesting" estimates:

```
# how many sperm whales at depths shallower than 2500m?
sum(predgrid$Nhat_tw[predgrid$Depth < 2500])

## [1] 1006.27

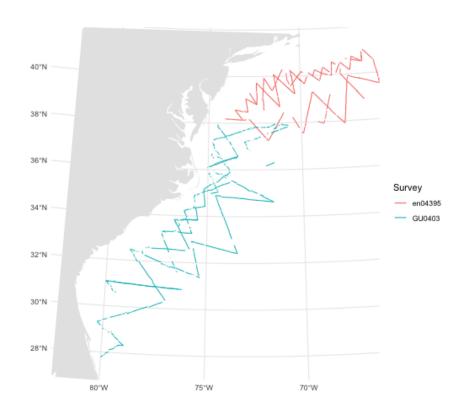
# how many sperm whales East of 0?
sum(predgrid$Nhat_tw[predgrid$x>0])

## [1] 1383.744
```

Extrapolation

What do we mean by extrapolation?

- Predicting at values outside those observed
- What does "outside" mean?
 - o between transects?
 - outside "survey area"?



Extrapolation

- In general, try not to do it!
- Variance issues?
- Space-time interchangability?
- dsmextra package by Phil Bouchet
 - https://densitymodelling.github.io/dsmextra/index.html



Prediction recap

- Using predict
- Getting "overall" abundance
- Subsetting
- Plotting in R
- Extrapolation (and its dangers)

Estimating variance

Now we can make predictions

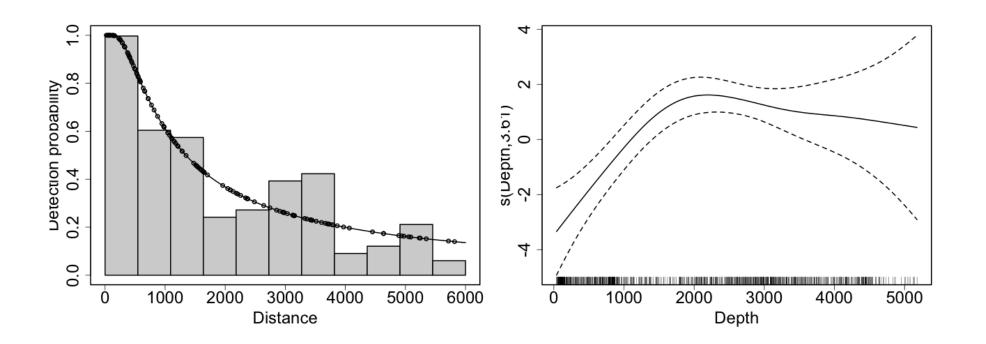
Now we are dangerous.

Predictions are useless without uncertainty

Where does uncertainty come from?

Sources of uncertainty

- Detection function parameters
- GAM parameters
- (And more! But only looking at these 2 here!)



Uncertianty of what?

- Uncertainty from detection function + GAM
- Want to talk about \hat{N} , so need to do some maths
- dsm does this for you!
- Details in Miller et al (2013) appendix

GAM + detection function uncertainty

(Getting a little fast-and-loose with the mathematics)

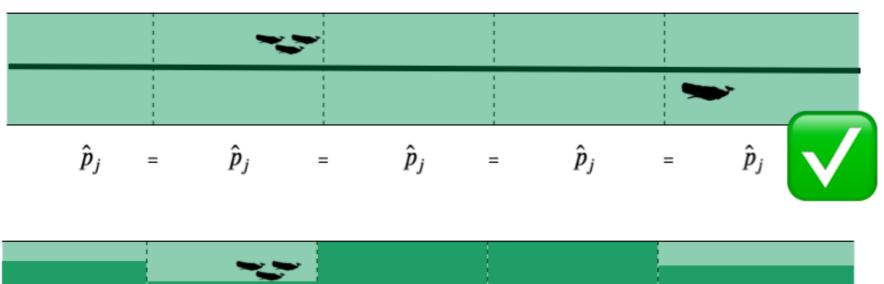
$$CV^2(\hat{N}) \approx CV^2(GAM) +$$

CV² (detection function)

the "delta method"

When can we use the delta method?

- Assumes detection function and GAM are independent
- This is okay if:
 - no detection function covariates



Variance propagation

- When detection function is not independent
- Uncertainty "propagated" through the model
- Refit both models together
- Bravington, Miller and Hedley (2019)
 - https://arxiv.org/abs/1807.07996

In R...

- Functions in dsm to do this
- dsm.var.gam
 - assumes spatial model and detection function are independent
- dsm.var.prop
 - propagates uncertainty from detection function to spatial model
 - only works for count models
 - covariates can only vary at segment level

Variance of abundance

Using dsm.var.gam

```
dsm_tw_var_ind <- dsm.var.gam(dsm_all_tw_rm, predgrid,</pre>
                              off.set=predgrid$off.set)
summary(dsm tw var ind)
  Summary of uncertainty in a density surface model calculated
##
   analytically for GAM, with delta method
##
  Approximate asymptotic confidence interval:
##
      2.5%
               Mean
                       97.5%
## 1539.017 2491.863 4034.641
## (Using log-Normal approximation)
##
## Point estimate
                                : 2491.863
## CV of detection function
                            : 0.2113123
## CV from GAM
                               : 0.1329
## Total standard error : 622.0386
## Total coefficient of variation: 0.2496
```

Plotting - data processing

- Calculate uncertainty per-cell
- dsm.var.* thinks predgrid is one "region"
- Need to split data into cells (using split())
- Need width and height of cells for plotting

Plotting (code)

##

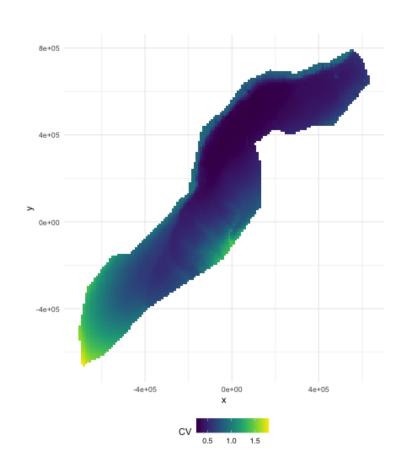
```
predgrid$width <- predgrid$height <- 10*1000</pre>
predgrid_split <- split(predgrid, 1:nrow(predgrid))</pre>
head(predgrid_split,3)
## $\1\
##
           x y Depth SST NPP DistToCAS
## 126 547984.6 788254 153.5983 12.04609 1462.521 11788.97
             EKE off.set long lat Nhat_tw
##
## 126 0.0008329031 1e+08 -66.52252 40.94697 0.01417646
##
     height width
## 126 10000 10000
##
## $`2`
##
                     Depth SST NPP DistToCAS
           Χ
                 V
## 127 557984.6 788254 552.3107 12.81379 1465.41 5697.248
             EKE off.set long lat Nhat_tw
##
height width
##
## 127 10000 10000
##
## $`3`
##
                     Depth SST
                                      NPP DistToCAS
                 V
           Χ
## 258 527984.6 778254 96.81992 12.90251 1429.432
                                           13722.63
```

EKE off.set long

lat

Nhat tw

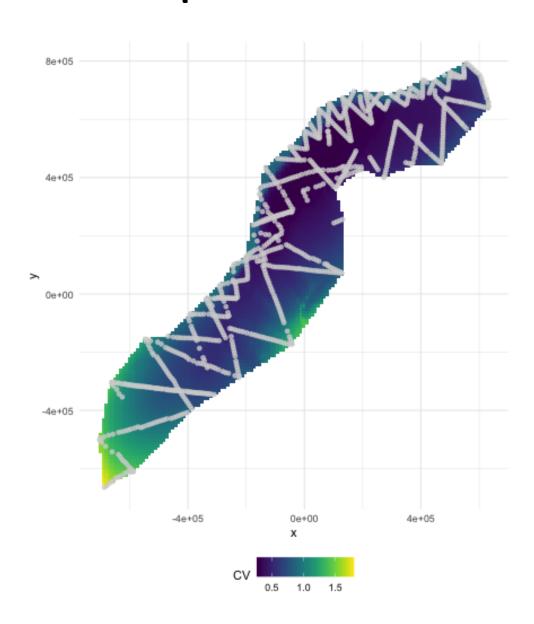
CV plot



Interpreting CV plots

- Plotting coefficient of variation
- Standardise standard deviation by mean
- $CV = se(\hat{N})/\hat{N}$ (per cell)
- Can be useful to overplot survey effort

Effort overplotted



Big CVs

- Here CVs are "well behaved"
- Not always the case (huge CVs possible)
- These can be a pain to plot
- Use cut() in R to make categorical variable
 - e.g. c(seq(0,1, len=10), 2:4, Inf) or somesuch
- (Example in practical)

Uncertainty recap

- How does uncertainty arise in a DSM?
- Estimate variance of abundance estimate
- Map coefficient of variation

Practical advice

Pilot studies and "you get what you pay for"

- Designing surveys is hard
- Designing surveys is essential

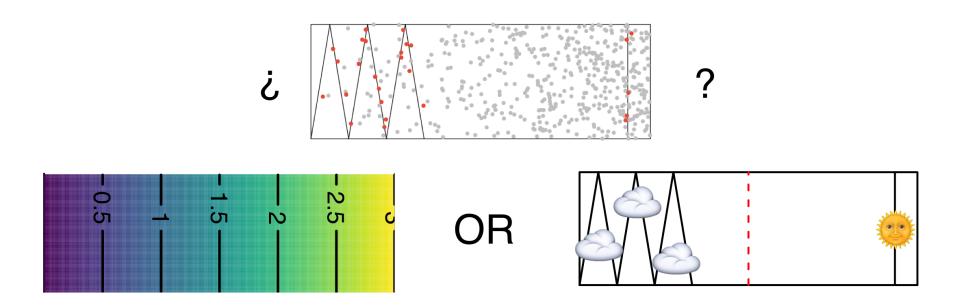
- Better to fail one season than fail for 5, 10 years
- Get information early, get it cheap
 - Inform design from a pilot study

Avoiding rules of thumb

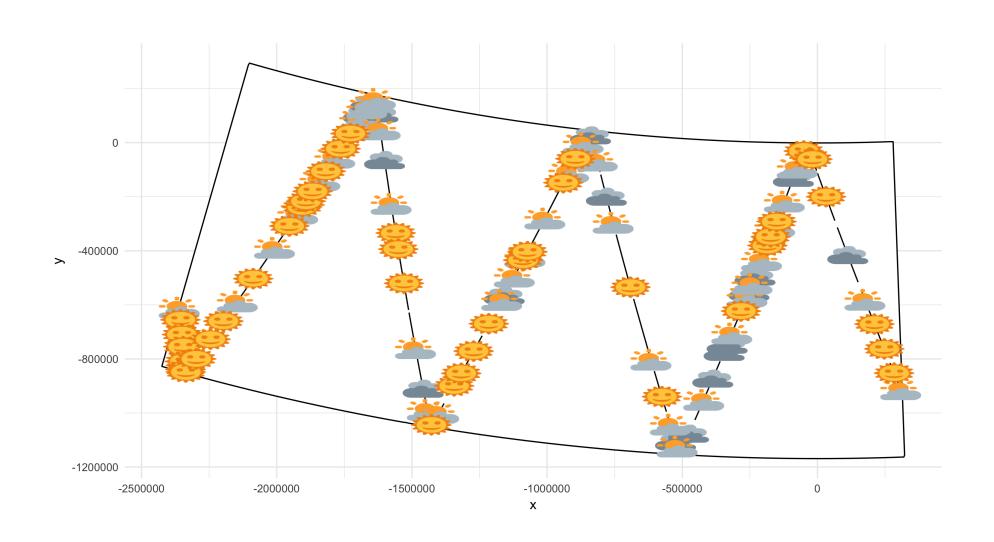
- Think about assumptions
 - Detection function
 - Spatial model
- Think about design
 - Spatial coverage
 - Covariate coverage

Sometimes things are complicated

- Weather has a big effect on detectability
- Need to record during survey
- Disambiguate between distribution/detectability
- Potential confounding can be BAD



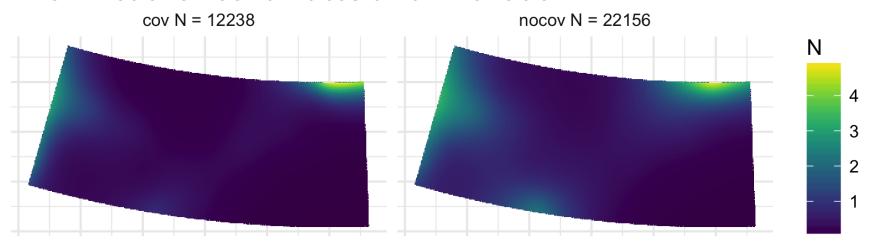
Visibility during POWER 2014



Thanks to Hiroto Murase and co. for this data!

Covariates can make a big difference!

- Same data, same spatial model
- With weather covariates and without



Disappointment

- Sometimes you don't have enough data
- Or, enough coverage
- Or, the right covariates

Sometimes, you can't build a spatial model

Segmenting

- Example on course site
- Length of $\approx 2w$ is reasonable
- Too big: no detail
- Too small: all 0/1
- See also Redfern et al., (2008)

Getting help

Resources

- Course reading list has pointers to these topics
- DenMod wiki with FAQ and more
- Distance sampling Google Group
 - Friendly, helpful, low traffic
 - see distancesampling.org/distancelist.html

That's all folks!