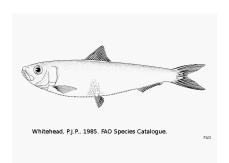


Assessment For All (a4a)

The stock assessment model



The a4a team
European Commission
Joint Research Centre







The a4a stock assessment framework

The a4a stock assessment framework is a statistical catch-at-age model written in R with ADMB underneath (to find the parameters). It was written with the aim of quickly developing stock assessment models for stocks with limited time series.





The a4a stock assessment framework

A model is defined by submodels, which specify the different parts of a statistical catch-at-age model.

There are 5 submodels in operation:

- · a model for F-at-age
- · a model for the initial age structure
- · a model for recruitment
- a (list) of model(s) for abundance indices catchability-at-age
- a list of models for the observation variance of catch-at-age and abundance indices.



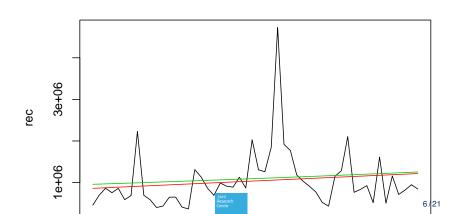


The a4a stock assessment framework

The submodels form use linear models. This opens the possibility of using the linear modelling tools available in R: see for example the mgcv gam formulas, or factorial design formulas using lm(). In R's linear modelling language, a constant model is coded as \sim 1, while a slope over age would simply be \sim age. For example, we can write a traditional year/age separable F model like \sim factor(age) + factor(year).



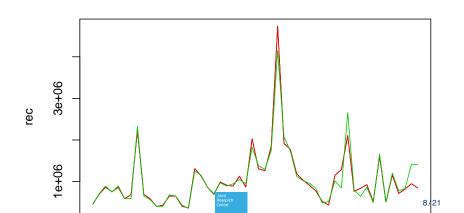






```
rlm <- lm(rec factor(year)) scafit <- sca(ple4, pl
srmodel = factor(year))</pre>
```







Model detail

$$e^{\mathsf{E}[\log C]} = \frac{\mathsf{F}}{\mathsf{F} + \mathsf{M}} (1 - e^{-\mathsf{F} - \mathsf{M}}) \mathsf{R} e^{-\sum \mathsf{F} + \mathsf{M}}$$

and

$$e^{E[\log I]} = ORe^{-\sum F + M}$$

and

$$Var[log C_{ay}] = \frac{\sigma_{ay}^2}{\sigma_{ays}^2}$$
 $Var[log I_{ays}] = \frac{\tau_{ays}^2}{\sigma_{ays}^2}$





Model detail

linear models for

- log F
- · log Q
- log observation variances
- · log initial age structure

Recruitment is modelled as a fixed variance random effect with linear models for

- · log a
- log b

where relevant. Models available: Ricker, Beverton Holt, smooth hockeystick, geometric mean





Linear models

It is not always obvious that stock assessments are often composed of linear models.

For example, the classical separable F assumption is simply that

$$F_{ay} = S_a \times F_y$$

which, in linear modelling parlance is

$$\log F \sim \text{age} + \text{year}$$



Intuitive Modelling

The "language" of linear models has been developing within the statistical community for many years:

- 1965 J. A. Nelder, notation for randomized block design
- 1973 Wilkinson and Rodgers, symbolic description for factorial designs
- · 1990 Hastie and Tibshirani, introduced notation for smoothers
- 1991 Chambers and Hastie, further developed for use in S

Many modelling software use this language: Minitab, spss, genstat, SAS, R, S-plus.



Some examples

A separable model where the level of F is smooth through time

$$\log F \sim \text{age} + \text{s(year)}$$



Some examples

A separable model where F is smooth over age

$$\log F \sim s(age) + year$$



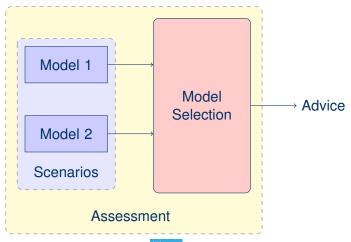
Some examples

F is smooth over age and year

 $\log F \sim s(age, year)$

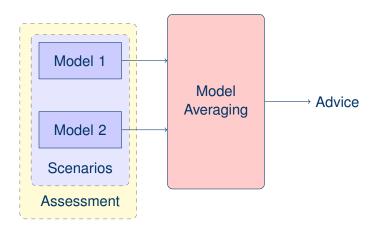


An Assessment Process





Model Averaging can help automation





Expert knowledge for model specification

Different plausible models for different levels

- · Management area level (North Sea, Baltic Sea, ...)
- Species type (roundfish, flatfish, pelagic, Nephrops)
- specific groups (North Sea gadoids)

This provides a framework for setting up plausible models for new species.

Can lots of simple models averaged = a good model?

Kearns: Can a set of weak learners create a single strong learner





Thank you for listening!





What we can do, what we can't do

Can:

- · missing values: missing at random
- · multiple surveys
- · variable Q, F, variance
- splines (fixed degreed of freedom)
- stock recruit relationship (fixed variance)
- stock recruit relationship (estimated variance) SLOW
- fixed variance random effects: RW1, RW2, seasonal, user specified

Can't:

- estimate random effect variance
- estimate smoothing parameters
- estimate growth parameters





What we can do

- simulate from the distribution of model params
 - · normal approx
 - · avoids the need for delta approx
 - · can be biased, but we can also use MCMC if desired
- · we can approximate the (joint) distribution of
 - · terminal year Fs and Ns
 - terminal year Fbar and Fmsy
 - F / Fmsy

