# Further Exploration of the Operating Model for the Management Procedure Evaluation. Hiroyuki Kurota <br> National Research Institution of Far Seas Fisheries 

Shimizu, Shizuoka, Japan


#### Abstract

Specification of the Operating Model for the Management Procedure development is further explored. First, results of the core set specified as the basis for the final reference set are examined. They are highly dependent on the CPUE series used and the pessimistic result is considered to be related to low estimate of the omega parameter (CPUE-abundance relationship). Several sensitivity analyses regarding the tagging data weight and CPUE assumptions are also conducted. Finally, alternative model specifications are tested. Error structure and sample size for age and length composition data influence conditioning and future projection results significantly. It is necessary to determine the final reference set and the robustness trials in consideration of these results.


## Introduction

The Stock Assessment Group (SAG) meeting in September 2004 determined that the Operating Model (OM) for the Management Procedure (MP) evaluation should be further developed, because all problems regarding the OM had not been solved in the meeting (CCSBT, 2004). Members are required to examine results of the core set specified as the basis for the final reference set and explore the impact of changes in the model assumptions and input data. This document refers to several important factors to determine the final reference set and the robustness trials, which will be finalized in February 2005 at the Seattle meeting.

The SAG adopted a new approach called the grid approach to integrate uncertainties of the OM into the reference set in place of the MCMC approach due to time constraints. The grid approach constructs the reference set from a large number of MPD results (scenarios) which cover the range of uncertainties in some fundamental parameters and input data: steepness, M0 (natural mortality at age 0), M10 (natural mortality over age 10), omega (CPUE-abundance relationship) and CPUE series, and the weight of scenarios is assigned based on priors (steepness and CPUE) or the combination of priors and likelihood (M0, M10 and omega). Following the grid approach, all analyses in this document were basically conducted with "main.tpl", "sbtmod14.tpl" and "sample.tpl", which were programs of the AD model builder (ADMB) distributed on 30 September 2004 by Ana Parma.

Future projections at the current catch ( 14930 mt ) were also calculated to investigate model behaviors. It took over five hours to complete one grid integration, although the calculation time depended on model assumptions and machine power of PC.

## Consideration of the core set

- general feature and future projection result

The core set is integrated across the five grid axes from 270 scenarios by the grid approach and it consists of 2000 stochastic realizations. Before examining results of the core set, I investigated numerical problems related to the local minima found for the previous versions (Panel, 2004). Several different initial values were set for the estimation. Converged estimates did not change from the originally estimated values in almost all cases where I had examined, although the maximum gradient component in convergence calculated by the ADMB was not so low (more than $10^{-4}$ ). The core set would be more robust to the problem related to the local minima.

Figure 1 shows historical estimates of biomass and recruitment and future projections at the current catch for the core set. The projection result became more pessimistic than those with the previous OMs and the population would collapse in 30 years in many cases. I also conducted projections using different constant catches (Figure 2, Table 1). This pessimistic result would influence our choice of tuning levels for the MP evaluation.

## - strong interaction of CPUE and omega

At first, I suspected that the pessimistic result of the core set was caused mainly by low recent recruitments in 2000 or later. However, population extinction was seen in many cases even without autocorrelation of future recruitments (noAC; Figure 3). This indicates that historical estimates of fundamental parameters of the population also become pessimistic.

The core set result was found to be highly dependent on CPUE series both in the conditioning and future projection (Figure 4). The result using the ST-window series was the most optimistic and biomass trend would be almost constant under the current catch. On the other hand, the w 0.5 series produced very pessimistic result and the population would collapse around 2015 under the current catches. I found that these differences reflected differences in M0, M10 and omega estimates which are closely connected to each of CPUE series. Table 2 shows the number of realizations for each value of M0, M10, and omega by different CPUE series. The comparison between the ST-window and the w 0.5 reveals a significant difference in distribution of omega values. The ST-window settled with the omega as 1.0 in much higher proportion than that for the w0.5. Figures 5 and 6 show conditioning results of "h2m2M2O2C4" (ST-window) and "h2m2M2O1C5" (w0.5), respectively,
each of which is one of major scenarios in each CPUE. Likelihood components of the two are different mainly in CPUE and Indonesia CAA fittings (Table 3).

In all CPUE series except the ST-window and the Laslett, the omega parameter put higher weight to 0.75 than 1.0 , although the prior weight of 0.75 was smaller (Table 2 ). I consider that this low omega is one of major reasons for the pessimistic result of the core set. It is necessary to pay more attention to omega values and their interaction to CPUE series as well as natural mortality.

- low recruitment in 2000 and 2001 with low CVs

Recruitment estimates before 2001 are used for the future projection without any modification. The narrow error bound on the recent recruitment estimates is one of issues to be further considered. Basson et al. (2004) indicated that CVs on recent recruitments in each scenario, which are approximated from the Hessian by the ADMB, are higher than those on past recruitments around 1960-1990. The same trend was observed when examining individual scenario independently as shown in Figure 7a as an example. This reflects limited amount of information on recent recruitment and it is quite natural.

However, the examination of CVs of recruitment obtained from all 2000 realizations of the core set (from Cfullnotag.s4) showed a different trend (Figure 7b). CVs were about 0.3 to 0.4 through all times except late 1970s and 1980s and they were almost constant from 1990 to 2001. This result shows that CVs in 2000 and 2001 are not lower compared to those in other years, but it also indicates that the time dependency is not strong. This difference in CV trends between each individual scenario and the core set must relate to the grid integration. If the grid approach is continued to be used, this issue would be needed for discussion.

I also investigated the reason for low recruitments in 2000 and 2001. In place of developing the program code for the retrospective analysis, I replaced the size composition data of LL1 catches in 2002 and 2003 with the average value of 1999-2001 data. This is an examination of what would be happened if no change were occurred in LL1 catch in 2002 and 2003, although there were few small fish in 2002 and 2003 in reality. Run using the new replaced data showed that recruitments were higher than those for the base case (core set) and that they were at almost the same level as those in the previous years (Figure 8). Therefore I consider that one of the reasons for the low recruitment estimate is lack of small fish in 2002 and 2003.

## Sensitivity test

- tagging data

Further examination of tagging data was considered as a high priority for the analyses (CCSBT, 2004). New tagging data updated by Australian scientists were used in this analysis, but in general the conditioning results did not change significantly. The following is a brief summary of the results when incorporating new tagging data (Cfulltag):

1) B0 was lower than that for the core set and the decline rate (B2004/B0) was lower (Figure 9). Future projection was also more optimistic and the biomass was almost constant under the current catch (Figure 10, Table 4).
2) The error bounds on recent recruitments were narrower than those for the core set (Figure 7c).
3) Conditioning and projection results were highly dependent on CPUE series as in the core set (Figure 11, Table 5). B0 estimates were different among them. The ST-window had the most optimistic result and the w 0.5 did the most pessimistic one. This difference is considered to reflect differences both in M10 and omega estimates.
4) Natural mortality was generally higher than that for the core set. Figures 12 and 13 are conditioning results of "h2m2M2O2C4_tag" and "h2m2M3O1C5_tag", each of which is one of major scenarios in each CPUE (also see Table 3). When M10 was higher, selectivity of the LL1 and the Indonesian fishery for older fish was higher. Comparison between "h2m2M2O1C5_notag" (Figure 6) and "h2m2M3O1C5_tag" (Figure 13) revealed that fitting to the tagging data was improved significantly when the weight for the tagging data increased. However, "h2m2M2O2C4_notag" (Figure 5) and "h2m2M2O2C4_tag" (Figure 12) did not show any major difference. Higher M0 and M10 values are preferred in all CPUE series except the ST-window (Table 5).
5) Omega values totally increased in all CPUE series compared to those for the core set. This is considered to be one of the reasons why the result is more optimistic when incorporating the tagging data.

As noted in previous meetings, one major problem in the way the model handles the tagging data is that reporting rates are assumed to be known (CCSBT, 2004). It is a key issue how the uncertainty could be incorporated to the model, if the tagging data were incorporated in the conditioning.

## - CPUE (median CPUE and age range for selectivity standardization)

As noted earlier in the consideration of the core set, selection of CPUE series influences both conditioning and projection results substantially. Examination using the median of the five CPUE series showed that the general trend was similar to that for the core set, where the five different CPUE series were used individually (Figure 14, Table 6). However, the range of estimates was smaller. Especially the lower bound of biomass estimates in 2000 or later was close to the median,
although the reason has not been clarified yet. The median CPUE could not cover enough uncertainties which the five different CPUE series involve.

I conducted the sensitivity analysis to the age range to standardize selectivity for CPUE predictions (Figure 15, Table 7). The age range was changed from 4-30 to 8-12. Results differed substantially, especially in selecting higher proportion of M10 and omega, and lower estimation of B0 and the current biomass. The future projection became much more pessimistic.

## - Indonesian fishery selectivity

When the Indonesian fishery selectivity was assumed as constant over age 18, B0 and the current biomass were estimated to be lower and the future projection was more pessimistic (Figures 16, 17). It should be noted that the likelihood was much worse than for the base (Table 8).

## Alternative model specifications

## - Sample size and error structure for age and length composition data

The core set adopts the multinominal distribution as an error structure for age and length composition data. The sample size of each fishery was determined by the Panel in July 2004 by taking square of the old reference set times five to reduce the weight to the age and length composition data. I explored other alternatives for reducing the sample size and other error structures. The current program (sbtmod14.tpl) has additional two options for the error structure. One is the robust normal likelihood (Fournier et al., 1998) and the other is the log-normal distribution with variances based on the multinominal assumption and an additive variance term to consider additional process error (I called this "lognormal plus" approach). The additional process error was assumed to be the same among all fisheries.

Table 9 shows a summary of trials and results. I found that conditioning and future projection results depended on the error structure and the sample size significantly (Figures 18-25, Tables 10-17). Especially, the lognormal plus approach showed substantially different results. Although I have not examined each realization due to time constraints, this issue should be further examined. The following is a brief summary:

1) When using the old reference set as sample sizes, natural mortality, especially M10, had higher proportion in larger values than those for the core set irrespective of the error structure. B0 became lower. The projection results were more optimistic.
2) When the sample size was set as the half of the old reference set, the result was similar to that for the core set.
3) Runs using the robust normal distribution showed similar results as the default multinominal distribution.
4) The lognormal distribution resulted in higher M0 and omega values and more optimistic results. In addition, as the process error was assumed to be larger, the omega estimates became larger.

## - age-dependent natural mortality

I explored different forms for age-dependency of natural mortality. Because the tagging data showed a strong interaction with natural mortality as noted earlier, the new modeling of natural mortality was applied both to "notag" and "tag" scenarios. Two alternatives were investigated to relax assumptions. First, the "m-slope", a curvilinear function to connect M0 and M10, was estimated instead of fixing it as 0.7 . The range of m-slope was constrained from 0.3 to 1.2 in the estimation. Estimates of the slope were often different from the default value, but general form of age-dependency was not different significantly regardless of whether or not the tagging data was involved (Figures 26,27) and the likelihood was not improved substantially enough to introduce the new parameter. Biomass and recruitment estimates also did not change (Figures 28, 29).

In some tuna species, natural mortality is considered to increase when fish is beyond a certain age (Hampton et al., 2003). If this is the case for SBT, it might influence not only natural mortality for young fish but also selectivity estimates. Therefore, in addition to M0 and M10, I introduced a new parameter "M20", which is natural mortality at age 20, by changing the program. I assumed that the mortality between age 10 and 20 changed linearly and it was constant beyond age 20. The range of M20 was set from 0.03 to 0.2 . The results showed that average M10 estimates among all realizations were lower than those for the core set and M20 estimates were higher slightly than M10 (Figure 30). Historical biomass estimates generally did not change, but the recent trend of biomass was different especially when including the tagging data (Figures 31, 32). However, since the current projection program did not meet mortality change over age 10, the projection results were not able to be obtained.

## Reference

Basson, M., Polacheck, T., Kolody, D., Preece A., and Hartog, J. 2004. Implications for management procedure evaluation: the mechanical update and further exploration of the operating model. CCSBT-ESC/0409/24.

CCSBT 2004. Report of the fifth meeting of the stock assessment group. Jeju, Republic of Korea, September 2004.

Fournier, D.A, Hampton, J., and Sibert, J.R. 1998. MULTIFAN-CL: a length-based, age-structured model for fisheries stock assessment, with applications to South Pacific albacore, Thunnus
alalunga. Can. J. Fish. Aquat. Sci. 55:1-12.
Hampton, J., Kleiber, P., Takeuchi, Y., Kurota, H., and Maunder, M. 2003. Stock assessment of bigeye tuna in the western and central Pacific Ocean, with comparisons to the entire Pacific Ocean. SCTB16 Working Paper BET-1.


Figure 1. 10th, median and 90th percentiles of historical estimates of biomass (upper panels; left 1931-2032, right 1980-2032) and recruitment (lower panels; left 1931-2032, right 1980-2032) and future projections using the current catch for the core set. "ALL" represents the median of all 2000 realizations.


Figure 2. Future projection for the core set at different constant catches.

Table 1. B2022/B2004 for the core set at different constant catches.

## Cfullnotag

| constant catch <br> level | B2022/B2004 |
| ---: | ---: |
| 14930 | 0.42 |
| 0 | 1.81 |
| 1000 | 1.74 |
| 2000 | 1.68 |
| 3000 | 1.62 |
| 4000 | 1.57 |
| 5000 | 1.50 |
| 6000 | 1.40 |
| 7000 | 1.29 |
| 8000 | 1.20 |
| 9000 | 1.11 |
| 10000 | 1.01 |
| 11000 | 0.88 |
| 12000 | 0.76 |
| 13000 | 0.64 |
| 14000 | 0.52 |
| 15000 | 0.42 |
| 16000 | 0.32 |
| 17000 | 0.21 |
| 18000 | 0.08 |
| 19000 | 0.01 |
| 2000 | 0.00 |



Figure 3. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch for the noAC.


Figure 4. Comparison of the median of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch for the core set among different CPUE series (ALL: all five CPUE series, CPUE2: nominal, CPUE3: Laslett, CPUE4: ST-window, CPUE5: w0.5, CPUE6: w0.8).

Table 2. The number of realizations in each level of M0, M10 and omega parameters in different CPUE series for the core set.

|  | $\mathrm{m0}$ |  |  | m 10 |  |  | omega |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| cpue | 0.3 | 0.4 | 0.5 | 0.07 | 0.1 | 0.13 | 0.75 | 1.0 |
| nominal (2) | 175 | 138 | 87 | 41 | 248 | 111 | 219 | 181 |
| Laslett (3) | 171 | 150 | 79 | 74 | 256 | 70 | 178 | 222 |
| st-window (4) | 169 | 138 | 93 | 88 | 202 | 110 | 48 | 352 |
| w0.5 (5) | 165 | 138 | 97 | 23 | 320 | 57 | 388 | 12 |
| w0.8 (6) | 165 | 144 | 91 | 48 | 282 | 70 | 376 | 24 |
| all | 845 | 708 | 447 | 274 | 1308 | 418 | 1209 | 791 |
| average | 0.380 |  |  | 0.102 |  |  | 0.849 |  |

Model: sbtmod14





TAG 1992



1993





$$
h=0.55, M 10=0.10, M 0=0.40
$$



Figure 5. Conditioning results of h2m2M2O2C4_notag (core set).


$$
h=0.55, M 10=0.10, M 0=0.40
$$



Figure 6. Conditioning results of h2m2M2O1C5_notag (core set).

Table 3. MPD results for some selected runs in the core set and the Cfulltag set.

(a) Example of h2m2M2O2C4_notag

(b) core set


(c) Cfulltag


Figure 7. CVs of estimated recruitments.


Figure 8. Biomass (left) and recruitment (right) estimates using catch-at-size data of LL1 in 2002 and 2003 replaced by the average of 1999-2001 data.


Figure 9. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch for the Cfulltag set.


Figure 10. Future projection for the Cfulltag set at different constant catches.

Table 4. B2022/B2004 for the Cfulltag at different constant catches.
Cfulltag

| Clonstant catch <br> level | B2022/B2004 |
| ---: | ---: |
| 14930 | 0.87 |
| 0 | 2.07 |
| 1000 | 2.01 |
| 2000 | 1.96 |
| 3000 | 1.90 |
| 4000 | 1.84 |
| 5000 | 1.78 |
| 6000 | 1.71 |
| 7000 | 1.63 |
| 8000 | 1.55 |
| 9000 | 1.47 |
| 10000 | 1.39 |
| 11000 | 1.27 |
| 12000 | 1.16 |
| 13000 | 1.07 |
| 14000 | 0.97 |
| 15000 | 0.87 |
| 16000 | 0.77 |
| 17000 | 0.68 |
| 18000 | 0.57 |
| 19000 | 0.48 |
| 20000 | 0.40 |



Figure 11. Comparison of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch for the Cfulltag set among different CPUE series.

Table 5. The number of realizations in each level of M0, M10 and omega parameters in different CPUE series for the Cfulltag set.

|  | m 0 |  |  | m 10 |  | omega |  |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| cpue | 0.3 | 0.4 | 0.5 | 0.07 | 0.1 | 0.13 | 0.75 | 1.0 |
| nominal (2) | 119 | 119 | 162 | 6 | 170 | 224 | 65 | 335 |
| Laslett (3) | 119 | 133 | 148 | 53 | 205 | 142 | 76 | 324 |
| st-window (4) | 114 | 170 | 116 | 75 | 228 | 97 | 20 | 380 |
| w0.5 (5) | 21 | 146 | 233 | 0 | 171 | 229 | 346 | 54 |
| w0.8 (6) | 82 | 107 | 211 | 2 | 229 | 169 | 311 | 89 |
| all | 455 | 675 | 870 | 136 | 1003 | 861 | 818 | 1182 |
| average | 0.421 |  |  | 0.111 |  |  | 0.898 |  |



$$
h=0.55, M 10=0.10, M 0=0.40
$$



Figure 12. Conditioning results of h2m2M2O2C4_tag (Cfulltag).


$$
h=0.55, M 10=0.14, M 0=0.40
$$



Figure 13. Conditioning results of h2m2M3O1C5_tag (Cfulltag).


Figure 14. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch for the sensitivity analysis using the median CPUE.

Table 6. The number of realizations in each level of M0, M10 and omega parameters for the sensitivity analysis using the median CPUE.

|  | $\mathrm{m0}$ |  |  | m 10 |  | omega |  |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| cpue median | 0.3 | 0.4 | 0.5 | 0.07 | 0.1 | 0.13 | 0.75 | 1.0 |
| all | 847 | 713 | 440 | 296 | 1339 | 365 | 1354 | 646 |
| average | 0.380 |  |  | 0.101 |  |  | 0.831 |  |



Figure 15. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch when for the sensitivity analysis to the age range for selectivity standardization for CPUE prediction (CPUE age range 8-12).

Table 7. The number of realizations in each level of M0, M10 and omega parameters for the sensitivity analysis to the age range for selectivity standardization for CPUE prediction (CPUE age range 8-12).

|  | $\mathrm{m0}$ |  |  | $\mathrm{m10}$ |  |  | omega |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| cpue median | 0.3 | 0.4 | 0.5 | 0.07 | 0.1 | 0.13 | 0.75 | 1.0 |
| all | 944 | 686 | 370 | 41 | 875 | 1084 | 57 | 1943 |
| average | 0.371 |  |  | 0.116 |  |  | 0.993 |  |

Model: sbtmod14








$$
h=0.55, M 10=0.14, M 0=0.40
$$



Figure 16. Conditioning results of h2m2M3O2 when Indonesia selectivity is constant over age 18.


Figure 17. Projection results of h 2 m 2 M 3 O 2 at the current catch when Indonesia selectivity is assumed to be constant over age 18.

Table 8. MPD results for sensitivity tests to Indonesian fishery selectivity.

|  |  | $\begin{aligned} & \left\lvert\, \begin{array}{l} \text { base } \\ \text { h1m2M3O2 } \end{array}\right. \\ & \hline \end{aligned}$ | $\begin{aligned} & \text { Indo18 } \\ & \text { h1m2M3O2 } \end{aligned}$ | base <br> h2m2M3O2 | $\begin{aligned} & \text { Indo18 } \\ & \text { h2m2M3O2 } \\ & \hline \end{aligned}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | Steepness | 0.385 | 0.385 | 0.55 | 0.55 |
|  | $\mathrm{M}(0)$ | 0.4 | 0.4 | 0.4 | 0.4 |
|  | M(10) | 0.14 | 0.14 | 0.14 | 0.14 |
|  | Omega | 1 | 1 | 1 | 1 |
|  | Cpue | 1 | 1 | 1 | 1 |
|  | Total | 464.007 | 485.839 | 464.211 | 476.824 |
| Likelihood | LL1 | 136.869 | 141.469 | 137.584 | 141.337 |
|  | LL2 | 48.6888 | 48.2111 | 48.7638 | 48.358 |
|  | LL3 | 108.155 | 105.794 | 107.221 | 105.403 |
|  | LL4 | 136.323 | 141.285 | 136.291 | 140.525 |
|  | IND | 25.1971 | 36.6995 | 24.7691 | 33.0298 |
|  | SURF | 32.0806 | 32.1604 | 32.0735 | 32.0478 |
|  | CPUE | -59.7215 | -62.361 | -59.0241 | -62.2476 |
|  | Tags | $6.32677 \mathrm{E}-05$ | 4.94273E-05 | 5.68439E-05 | $5.40181 \mathrm{E}-05$ |
| Penalties | Sel.Ch | 36.7591 | 35.9219 | 36.0817 | 35.0044 |
|  | Sel.sm | 20.6836 | 20.9524 | 20.5111 | 20.6734 |
|  | Sg.R | -21.0275 | -14.2932 | -20.0598 | -17.3067 |
|  | $\mathrm{M}(0)$ | 0 | 0 | 0 | 0 |
|  | M(10) | 0 | 0 | 0 | 0 |
|  | Steepness | 0 | 0 | 0 | 0 |
| Ref. Pts | msy | 21579.5 | 18138 | 27230 | 24538.6 |
|  | S(msy) | 356101 | 299277 | 221967 | 200066 |
|  | S(msy)/B0 | 0.41688 | 0.417168 | 0.329563 | 0.329864 |
|  | S(2004)/S(0) | 0.134441 | 0.083325 | 0.140363 | 0.081919 |
|  | S(2004) | 114840 | 59777.5 | 94537.4 | 49684.7 |
| Rho | 1931-Y | 0.631824 | 0.718777 | 0.655112 | 0.707721 |
|  | 1965-1998 | 0.447061 | 0.351059 | 0.446395 | 0.465627 |
| SigmaR | Model SigR | 0.6 | 0.6 | 0.6 | 0.6 |
|  | 1931-Y | 0.405374 | 0.458608 | 0.41777 | 0.447453 |
|  | 1965-1998 | 0.308475 | 0.285441 | 0.303347 | 0.306416 |
| CPUE | 1969-Y | 0.121398 | 0.0566797 | 0.0966185 | 0.0232437 |
| Autocorr. | 1990-2000 | 0.383489 | 0.339954 | 0.32519 | 0.281126 |

Table 9. A summary of trials and results on the error structure and the sample size for age and length composition data.

| option | Fig | Table | error structure | sample size* | M0 | M10 | omega | projection <br> (median of B2032) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| base | 3 | 2 | multinominal | core set | 0.380 | 0.102 | 0.849 | extinction |
| a | 18 | 10 | multinominal | old reference set | 0.404 | 0.122 | 0.859 | almost current level |
| b | 19 | 11 | multinominal | old reference set / 2 | 0.381 | 0.114 | 0.837 | extinction |
| c | 20 | 12 | robust normal | old reference set | 0.406 | 0.122 | 0.849 | slight decline trend |
| d | 21 | 13 | robust normal | core set | 0.400 | 0.109 | 0.868 | decline trend |
| e | 22 | 14 | lognormal plus | old reference set + process error (0.0) | 0.377 | 0.126 | 0.887 | almost current level |
| f | 23 | 15 | lognormal plus | core set + process error (0.0) | 0.365 | 0.110 | 0.910 | slight decline trend |
| g | 24 | 16 | lognormal plus | core set + process error (0.05) | 0.370 | 0.110 | 0.922 | slight decline trend |
| h | 25 | 17 | lognormal plus | core set + process error (0.2) | 0.366 | 0.114 | 0.955 | almost current level |

* core set $=5 \times($ old reference set $\wedge 0.5)$


Figure 18. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch when error structure for age- and size-composition data is multinominal and sample size is old reference set (option a).

Table 10. The number of realizations in each level of M0, M10 and omega parameters in different CPUE series when error structure for age- and size-composition data is multinominal and sample size is old reference set (option a).

|  | $\mathrm{m0}$ |  |  | m 10 |  | omega |  |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| cpue | 0.3 | 0.4 | 0.5 | 0.07 | 0.1 | 0.13 | 0.75 | 1.0 |
| nominal (2) | 106 | 109 | 185 | 1 | 120 | 279 | 205 | 195 |
| Laslett (3) | 122 | 164 | 114 | 3 | 83 | 314 | 106 | 294 |
| st-window (4) | 127 | 167 | 106 | 1 | 69 | 330 | 80 | 320 |
| w0.5 (5) | 112 | 126 | 162 | 1 | 157 | 242 | 374 | 26 |
| w0.8(6) | 142 | 127 | 131 | 1 | 78 | 321 | 363 | 37 |
| all | 609 | 693 | 698 | 7 | 507 | 1486 | 1128 | 872 |
| average | 0.404 |  |  | 0.122 |  |  | 0.859 |  |



Figure 19. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch when error structure for age- and size-composition data is multinominal and sample size is old reference set / 2 (option b).

Table 11. The number of realizations in each level of M0, M10 and omega parameters in different CPUE series when error structure for age- and size-composition data is multinominal and sample size is old reference set / 2 (option b).

|  | $\mathrm{m0}$ |  |  | m 10 |  | omega |  |  |
| :---: | ---: | :---: | ---: | ---: | ---: | ---: | ---: | ---: |
| cpue | 0.3 | 0.4 | 0.5 | 0.07 | 0.1 | 0.13 | 0.75 | 1.0 |
| nominal (2) | 167 | 138 | 95 | 6 | 219 | 175 | 283 | 117 |
| Laslett (3) | 180 | 139 | 81 | 15 | 188 | 197 | 215 | 185 |
| st-window (4) | 170 | 141 | 89 | 14 | 151 | 235 | 64 | 336 |
| w0.5 (5) | 163 | 126 | 111 | 2 | 191 | 207 | 384 | 16 |
| w0.8 (6) | 165 | 143 | 92 | 5 | 210 | 185 | 355 | 45 |
| all | 845 | 687 | 468 | 42 | 959 | 999 | 1301 | 699 |
| average | 0.381 |  |  | 0.114 |  |  | 0.837 |  |



Figure 20. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch when error structure for age- and size-composition data is robust normal and sample size is old reference set (option c).

Table 12. The number of realizations in each level of M0, M10 and omega parameters in different CPUE series when error structure for age- and size-composition data is robust normal and sample size is old reference set (option c).

|  | $\mathrm{m0}$ |  |  | m 10 |  |  | omega |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| cpue | 0.3 | 0.4 | 0.5 | 0.07 | 0.1 | 0.13 | 0.75 | 1.0 |
| nominal (2) | 113 | 160 | 127 | 23 | 64 | 313 | 167 | 233 |
| Laslett (3) | 75 | 231 | 94 | 18 | 93 | 289 | 183 | 217 |
| st-window (4) | 61 | 196 | 143 | 9 | 70 | 321 | 111 | 289 |
| w0.5 (5) | 91 | 151 | 158 | 1 | 74 | 325 | 391 | 9 |
| w0.8 (6) | 157 | 143 | 100 | 27 | 93 | 280 | 358 | 42 |
| all | 497 | 881 | 622 | 78 | 394 | 1528 | 1210 | 790 |
| average | 0.406 |  |  | 0.122 |  |  | 0.849 |  |



Figure 21. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch when error structure for age- and size-composition data is robust normal and sample size is core set (option d).

Table 13. The number of realizations in each level of M0, M10 and omega parameters in different CPUE series when error structure for age- and size-composition data is robust normal and sample size is core set (option d).

|  | $\mathrm{m0}$ |  |  | m 10 |  |  | omega |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| cpue | 0.3 | 0.4 | 0.5 | 0.07 | 0.1 | 0.13 | 0.75 | 1.0 |
| nominal (2) | 174 | 126 | 100 | 69 | 133 | 198 | 199 | 201 |
| Laslett (3) | 113 | 173 | 114 | 54 | 216 | 130 | 126 | 274 |
| st-window (4) | 69 | 213 | 118 | 17 | 285 | 98 | 32 | 368 |
| w0.5 (5) | 70 | 155 | 175 | 40 | 154 | 206 | 356 | 44 |
| w0.8 (6) | 183 | 106 | 111 | 88 | 85 | 227 | 346 | 54 |
| all | 609 | 773 | 618 | 268 | 873 | 859 | 1059 | 941 |
| average | 0.400 |  |  | 0.109 |  |  | 0.868 |  |



Figure 22. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch when error structure for age- and size-composition data is lognormal plus and sample size is old reference set (option e).

Table 14. The number of realizations in each level of M0, M10 and omega parameters in different CPUE series when error structure for age- and size-composition data is lognormal plus and sample size is old reference set (option e).

|  | $\mathrm{m0}$ |  |  | m 10 |  |  | omega |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| cpue | 0.3 | 0.4 | 0.5 | 0.07 | 0.1 | 0.13 | 0.75 | 1.0 |
| nominal (2) | 122 | 127 | 151 | 2 | 20 | 378 | 114 | 286 |
| Laslett (3) | 189 | 129 | 82 | 1 | 52 | 347 | 89 | 311 |
| st-window (4) | 195 | 143 | 62 | 0 | 33 | 367 | 61 | 339 |
| w0.5 (5) | 217 | 101 | 82 | 0 | 142 | 258 | 328 | 72 |
| w0.8 (6) | 191 | 125 | 84 | 0 | 32 | 368 | 314 | 86 |
| all | 914 | 625 | 461 | 3 | 279 | 1718 | 906 | 1094 |
| average | 0.377 |  |  | 0.126 |  |  | 0.887 |  |



Figure 23. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch when error structure for age- and size-composition data is lognormal plus and sample size is core set (option f ).

Table 15. The number of realizations in each level of M0, M10 and omega parameters in different CPUE series when error structure for age- and size-composition data is lognormal plus and sample size is core set (option f).

|  | $\mathrm{m0}$ |  |  | m 10 |  |  | omega |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| cpue | 0.3 | 0.4 | 0.5 | 0.07 | 0.1 | 0.13 | 0.75 | 1.0 |
| nominal (2) | 176 | 142 | 82 | 20 | 131 | 249 | 41 | 359 |
| Laslett (3) | 207 | 129 | 64 | 40 | 230 | 130 | 53 | 347 |
| st-window (4) | 211 | 141 | 48 | 45 | 241 | 114 | 21 | 379 |
| w0.5 (5) | 198 | 138 | 64 | 7 | 230 | 163 | 327 | 73 |
| w0.8 (6) | 218 | 121 | 61 | 22 | 238 | 140 | 275 | 125 |
| all | 1010 | 671 | 319 | 134 | 1070 | 796 | 717 | 1283 |
| average | 0.365 |  |  | 0.110 |  |  | 0.910 |  |



Figure 24. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections when error structure for age- and size-composition data is lognormal plus, sample size is core set and process error is 0.05 (option g).

Table 16. The number of realizations in each level of M0, M10 and omega parameters in different CPUE series when error structure for age- and size-composition data is lognormal plus, sample size is core set and process error is 0.05 (option g).

|  | $\mathrm{m0}$ |  |  | m 10 |  |  | omega |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| cpue | 0.3 | 0.4 | 0.5 | 0.07 | 0.1 | 0.13 | 0.75 | 1.0 |
| nominal (2) | 177 | 139 | 84 | 13 | 117 | 270 | 20 | 380 |
| Laslett (3) | 179 | 151 | 70 | 34 | 246 | 120 | 46 | 354 |
| st-window (4) | 192 | 153 | 55 | 37 | 262 | 101 | 14 | 386 |
| w0.5 (5) | 195 | 145 | 60 | 7 | 231 | 162 | 290 | 110 |
| w0.8(6) | 189 | 141 | 70 | 29 | 245 | 126 | 251 | 149 |
| all | 932 | 729 | 339 | 120 | 1101 | 779 | 621 | 1379 |
| average | 0.370 |  |  | 0.110 |  |  | 0.922 |  |



Figure 25. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections when error structure for age- and size-composition data is lognormal plus, sample size is core set and process error is 0.2 (option h).

Table 17. The number of realizations in each level of M0, M10 and omega parameters in different CPUE series when error structure for age- and size-composition data is lognormal plus, sample size is core set and process error is 0.2 (option h).

|  | $\mathrm{m0}$ |  |  | m 10 |  |  | omega |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| cpue | 0.3 | 0.4 | 0.5 | 0.07 | 0.1 | 0.13 | 0.75 | 1.0 |
| nominal (2) | 176 | 136 | 88 | 10 | 94 | 296 | 3 | 397 |
| Laslett (3) | 201 | 131 | 68 | 36 | 199 | 165 | 22 | 378 |
| st-window (4) | 217 | 137 | 46 | 30 | 219 | 151 | 3 | 397 |
| w0.5 (5) | 198 | 133 | 69 | 5 | 192 | 203 | 186 | 214 |
| w0.8 (6) | 226 | 111 | 63 | 19 | 192 | 189 | 144 | 256 |
| ALL | 1018 | 648 | 334 | 100 | 896 | 1004 | 358 | 1642 |
| average | 0.366 |  |  | 0.114 |  |  | 0.955 |  |



Figure 26. Average natural mortality as default cases (left: core set, right: tag).


Figure 27. Average natural mortality with 100 representative examples when mslope is estimated (left: notag, right: tag).


Figure 28. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch when m-slope is estimated (notag).


Figure 29. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch when m-slope is estimated (tag).


Figure 30. Average natural mortality with 100 representative examples when M20 is introduced (left: notag, right: tag).


Figure 31. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch when M20 is estimated (notag).


Figure 32. 10th, median and 90th percentiles of historical estimates of biomass (upper panels) and recruitment (lower panels) and future projections using the current catch when M20 is estimated (tag).

