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BRIEF REVIEW OF THE METHODS FOR THE FUTURE PROJECTIONS OF PACIFIC BLUEFIN TUNA STOCK ASSESSMENT

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Introduction

Future projections from estimates of stock assessment models can provide important information on the current and near-future status of concerned fish stocks, which should influence harvest strategy and help to maintain sustainable fisheries. As modern quantitative methods for stock assessment have became highly complex and computationally intensive, methods for future projections are becoming complex and computer intensive, as well. Integrated models with non-linear function optimizers and many parameters, such as Stock Synthesis II (SSII, Methot 2007) and Multifan-CL (Founier 1999), can estimate parameter uncertainty that accounts for much of the variability in the biological and fishery data. The uncertainty estimated from an integrated model should be carried forward into the projections, but straightforward methods to incorporate parameter uncertainty in projections, such as MCMC, require lengthy run times even when using state-of-the-art computer resources (Maunder et al. 2006).

Maunder et al. (2006) introduced a less computationally intense method to estimate uncertainty in projections that included the uncertainties in both future population demographics and parameter estimation. The future confidence intervals including both uncertainties were estimated by normal approximation based on the delta method (Oehlert 1992). In application to simulated population data and yellowfin tuna in the Eastern Pacific, the method showed reasonable performance when compared with the more computer intensive methods of MCMC and stochastic projection from point estimates (Maunder et al., 2006). Projections in SS II appear to be conducted using similar methods, although the SS II manuals (Methot 2005 and Methot 2007) do not provide a detailed description of its projection methodology.

On the other hand, stochastic, computationally-intensive statistical methods for more rigorously estimating uncertainty in projections (e.g. nonparametric bootstrapping) have a long history of application in fisheries stock assessment when simpler assessment methods are employed (e.g. These stochastic methods not only account for both parameter and demographic VPA). uncertainty but do so in manner that (i) makes no distributional assumptions about the parameters (e.g. normal, lognormal, etc.) and (ii) runs relatively quickly on modern computers. For example, recent ISC assessments have used such methods for future projections of North Pacific albacore (Conser et al. 2006) and Pacific bluefin tuna in the last stock assessment (Yamada et al. 2006). In April 2006, the Pacific bluefin tuna (PBF) working group agreed to use SS II rather than tuned VPA in the next stock assessment of Pacific bluefin tuna. Because the next stock assessment will employ more complex methodology for determining current stock size and the historical trend (with concomitant lengthy run times), it is important to evaluate the potential impacts on PBF stock size and demographic projections and the associated uncertainty. This is particularly relevant since ISC fishery managers have been relying on stochastic projection results to form the basis of fishery management. This document compares the results of future projections produced by SS II with those produced by more state-of-the-art stochastic projection methods. The results should provide background materials for working group discussion on the requirements for PBF projections that will be conducted in the next stock assessment.

Materials and methods

Sample data

We used fishery and biological data provided by the last working group (ISC PBF-WG/07-1) as sample data for SS II and future projections. Fishery and biological data for SS II were similar to that presented by Takeuchi et al. (2007) at the last working group, and model configurations were same as that by Kai et al. (2007, ISC PBF-WG/07-3/22). Because SS II was recently updated (v1.23e to v2.00g), some settings have been changed or added. Note that these PBF sample data and the SS II configurations used here are already out of date since the PBT database has been undergoing major reconstruction in preparation for the next stock assessment. Consequently, the results in this paper should only be used to examine competing projection methods but not to infer conclusions regarding the current stock status or future condition of Pacific bluefin tuna. The revised fishery and biological data as well as new potential SS II configurations will be presented in this and next meeting.

Stock evaluation and future projections by SS II were conducted using the sample data with iterative-reweighting (eight or more times) of effective sample size (McAllister and Ianelli 1997) and replacement of the SD associate with the respective CPUE indices. Each SS II run was conducted with the optional argument of '-maxfn 500' for the purpose of model stabilization. The sample data for the estimation of stock status included fishing years 1952-2004. Projections started in 2005 and continued through 2044 (40 years). Because steepness in Beverton-Holt spawner recruitment relationship was estimated as 1 in test runs, the default steepness was fixed as 1 for all runs in this paper. Further details on the settings for SS II are shown in Appendix B. The configuration described in this document is one of the most stable versions among all those produced from our experience of applying SS II to PBF. Although the sample case presented in this document can produce a positive definite Hessian matrix, most other trials (not shown here) failed to produce an inversed Hessian matrix and in many cases, or failed to even satisfy the standard convergence criteria. In addition, for a particular input option (relative F flag=2), calculation of Hessian matrix seemed to be especially unstable. In the bootstrap analysis, 44 of the 150 bootstrap replications failed to be converged, even for the most stable configuration. Finding a reasonable configuration for which parameter uncertainty can be properly estimated might be difficult task for PBF stock assessment itself (i.e. prior to conducting future projections).

Fishing scenarios for future projections with sample data

We conducted projections with three constant harvest rate scenarios: (1) SPR 0.1, (2) no-fishing, and (3) terminal year F. SPR 0.1 is the scenario where future constant harvest rates achieve future equilibrium SPR of 0.1. In this scenario, SS II projections were conducted using the harvest rates by fleet and season estimated in the terminal year (2004), and multiplied by various F multipliers. The SS II results from Scenario 2 (no fishing) were used to validate a set of stochastic projections functions written in the "R" programming language. In Scenario 3 (terminal year F), all future harvest rates were set at the 2004 level. The input settings for the SS II projections are shown in Tables 1-3 for Scenarios 1-3, respectively.

Characteristics of the projection method in SS II

As generally applied, SS II projections provide point estimates and standard deviations of the future population demographics of interest (such as recruitment and SSB) by inverting the Hessian matrix and applying the delta method¹. In this SS II mode, a projection is simply treated as part of the estimation model rather than as a separate stochastic process based on Monte Carlo simulation, e.g. as in Conser et al. (2005) or Brodziak (2005). When using the delta method, estimation of the standard deviations assumes that the underlying distribution is the normal (Gaussian). Therefore, it is noteworthy that the estimated parameters and derived quantities from the projection (e.g. future recruitment and SSB) are assumed to be normally distributed, with parameter estimates corresponding to 'arithmetic average (or expected value)', not with median as is the case for the lognormally distribution. It should be noted that recruitment deviation was assumed to be log-normal distribution in the historical phase in SS II (with log-bias adjustment factor), but normal in the projection phase (see input file settings in 4th, 5th lines at forecast.ss2 in Table 1).

Stochastic projections with R

Demographic stochastic projections were conducted using a program coded by R (R Development Core Team 2006). The program is an improved version of that used in the previous stock assessment of Pacific bluefin tuna (Yamada et al., 2006). The code was adapted to correspond to the SS II time steps (4 quarter per year). Pseudo code of the R program is shown in Appendix A. The characteristics of the R program for stochastic projections are listed as follows.

- Because steepness was fixed as 1 in SS II, future recruitment was assumed to be occurred with random log-normal distributions (eq. 1 in Appx A).
- Stochastic simulations were conducted with 1000 replications. The median, mean and probability distribution were calculated for recruitment, SSB, etc.

• While the population dynamics model used in SS II is based on Pope's approximate equation (in the current configuration of the sample data) with estimating vulnerable biomass and exploitation rates by fleets, stochastic projections used in this document were based on the catch equation with continuous F at age (eq. 2 and 3 in appx A). F at age used in the future projections was calculated from estimated population numbers and catch at age in a 2004 (eq. 4 in appx A).

• Stochastic projections from multiple SS II bootstrap estimations were attempted.

• Random re-sampling from recruitments estimated in the historical period was also carried out. Random re-sampling of historical recruitments can inhibit extraordinarily high or low recruitments compared with the historical mean recruitment of PBF. The previous assessment of PBF used re-sampling of historical recruitments for future recruitment

¹ Since SS II is written in AD Model Builder (ADMB), it is possible to use the ADMB MCMC functions in conjunction with SS II projections. While this is an ideal way to characterize the uncertainty in projection results (as well as terminal year results), MCMC is computationally intensive, requires considerable computer memory, and results in lengthy run times (often several days). In practice, MCMC is often not practical for SS II models with the complexity likely needed for the PBT assessment.

Calculation of reference point

As an example of a reference point based on stochastic projections, we calculated the probability that the SSB will fall below the historical lowest SSB during one or more years of the projection period. The definition of this probability is following Conser et al. (2005) (eq. 1). $\Pr[SSB_{Future} < SSB_{observed} | F] = \frac{1}{1000} \sum_{k=1}^{\infty} \left\| \min\{SSB_{2005,F}^{k}, ..., SSB_{2044}^{k}\} < \min\{SSB_{1952}, ..., SSB_{2004}\} \right\|$ (1), where the double bracket $\| \| \|$ indicates a logical test with outcome 0 (if false) or 1 (if true). We also consider another probability that SSB in 2040 (when SSB will be in equilibrium) will fall below the historical lowest SSB.

$$\Pr[SSB_{2040} < SSB_{observed} | F] = \frac{1}{1000} \sum_{k=1}^{1000} \left\| SSB_{2040}^k < Min\{SSB_{1952}, ..., SSB_{2004}\} \right\|$$
(2)

From the SS II projection results, we calculate the following probabilities:

$$\Pr[SSB_{future} < SSB_{observed} | F] = \max \begin{cases} \Pr[N(\mu_{2005}, \sigma_{2005}) < \min\{SSB_{1952}, ..., SSB_{2004}\}], \\ \Pr[N(\mu_{2006}, \sigma_{2006}) < \min\{SSB_{1952}, ..., SSB_{2004}\}], \\ \Pr[N(\mu_{2044}, \sigma_{2044}) < \min\{SSB_{1952}, ..., SSB_{2004}\}] \end{cases} \end{cases}$$
(3)

$$\Pr[SSB_{2040} < SSB_{observed} | F] = \Pr[N(\mu_{2040}, \sigma_{2040}) < \min\{SSB_{1952}, \dots, SSB_{2004}\}]$$
(4)

The probabilities defined in Eq. 2 and Eq 4 should be similar, but probabilities by eq. 1 and eq. 3 are not equivalent in the strict sense.

Results and discussion

Recruitment & numbers at age

Fig. 1 shows historical recruitments estimated with SS II, and projected recruitments with SS II and stochastic projections. Estimation of future recruitments from SS II (black lines) is equal to R_0 , and corresponds to the (simple) arithmetic average of the future recruitments projected by stochastic projections. On the other hand, median of the stochastic projections is less than R_0 because of the lognormal assumption for future recruitments, i.e. median = $R_0 * \exp(-\sigma/2)$. According to the normal approximation used in SS II (delta method), the 90% confidence interval of the future recruitments should be $R_0 \pm 1.64*$ (standard deviation of future recruitment). However, this resulted in negative recruitment values (see Fig 1b). Comparison of probability distributions of future recruitments in 2040 (Fig. 1b) shows clearly the difference of the two assumptions of future recruitments in SS II and Monte-Carlo based stochastic projections.

Numbers at age under SPR 0.1 scenario (fig. 2a) got similar outcomes with fig. 1a: simple average of numbers at age from stochastic projections and estimates values from SS II are corresponding each other, and median from stochastic projections is lower than that (except for the case of 20+). In the plus group, numbers at age from SS II were neither equal to average nor

median from stochastic projections. Considering the observation that average of the numbers at 20+ from stochastic projections is equal to SS II results under no-fishing scenario (fig. 2b), the discrepancy of 20+ projections between stochastic and SS II would caused by different catch equations used in the two method, and wrong way for calculating F for 20+ in our R code. This problem will be solved by improvement of our R code although the effect of the bias on the total SSB would be small because of small number of 20+ group.

Spawning biomass

Trajectories of SSB are compared in fig 3 and 4, under the scenario of no-fishing and SPR0.1, respectively. Estimated SSB by SS II and average ones by stochastic projections show similar results in the both scenarios, while probability distribution at 2040 (fig. 3b and 4b) and their 5%, 50%, and 95% percentiles were different. Probability distributions of SSB at 2040 from stochastic projections were skewed to lower SSB compared with approximated normal distribution in SS II, especially in SPR 0.01 scenario. In the near future period during which demographic stochastic projections were zero. On contrary, SS II estimated some confidence intervals more or less in the period because SS II future projection can consider parameter uncertainty as well as demographic stochastic uncertainty by using normal approximation.

Parametric bootstrap results of total exploitation rates shows parameter uncertainty of total exploitation rates (fig. 5a). Total number of reputation of bootstrap was 150 times, which took about 40 hours, and number of successful convergence was 106. The uncertainly was especially large during the last year of the assessment phase, when total exploitation rates ranged from 0.3 to 0.7 (fig. 5b).

The parameter uncertainty estimated from the bootstrap results was incorporated into stochastic projections by starting projections from the 106 bootstrap results with good convergence. Fig. 6 compares confidence intervals estimated from bootstrap+stochastic and SS II projections in the case of SPR0.1. Replication number of projections per each bootstrap result was 37 so that total number of replication was approximately 4000. Confidence interval estimated from the stochastic+bootstrap method was relatively skewed to higher SSB, and average value was not equivalent with the results from SS II. Possible reason of the higher SSB in the stochastic+bootstrap projections would be that point estimation of the total exploitation rate in 2004 was larger than average value of multiple bootstrap results (fig. 5b), which might suggest possible bias of total exploitation rate in. Consequently stochastic+bootstrap projections got more optimistic confidence interval near the future.

Projected SSB was nearly crashed both in the stochastic and SS II methods in the scenario of ending year F (fig. 7). This too pessimistic future perspective was caused from high total exploitation rates estimated in 2004. Fig. 8 shows results of retrospective analysis of total exploitation rates, and suggested that total exploitation rates of 0.55 in 2004 was the highest level

among those during 1990-2003 (0.21-0.35). Fig. 8 also suggested the possibility that exploitation rates at the end year have tendency of positive bias. Bootstrap results also show large uncertainty in total exploitation rates in the end year of 2004 (fig. 5b).

Future recruitment by resampling of historical recruitments

Fig. 9 shows other results based on random draw of future recruitments from historical recruitments under the scenario of SPR0.1, compared with the assumption of lognormal recruitment. Both median and 5% lower limit of recruitments with different assumptions was coincident each other, but average and 5% upper limit in resampling scenario were lower than those in lognormal scenario. As for SSB, future estimated median of SSB by resampling was also below that by lognormal while 5% lower limits was not different. This result suggest that too much high recruitments might overestimate future.

Example calculation of reference point

Patterns of Pr[SSB_{future}<SSB_{observed}] and Pr[SSB₂₀₄₀<SSB_{observed}] by fishing scenarios were different among different projection methods (Table 5). The percentages of Pr[SSB_{future}<SSB_{observed}] by SS II was higher than that by stochastic projections in the smaller F multiplier, but lower in the larger F multiplier. In particular, Pr[SSB_{future}<SSB_{observed}] by SS II was not zero even in SPR 0.1 scenario or the too precautionary scenario of no-fishing. Fig. 4 suggests that the lower 5% limit of SSB by SS II was below historical lowest SSB level during 2010-2014. According to the confidence interval from stochastic projections without parameter uncertainty, 'parameter uncertainty' estimated in SS II result in the lower 5% limit below historical lowest SSB during the period. It is doubtful that the estimated confidence interval by SS II is reasonably precautionary considering appropriate parameter uncertainty in this sample case since lower 5% limit estimated from bootstrap+stochastic projections was above that estimated by SSII (fig. 6).

Points for discussion

Definition of current F

In the default mode of SS II, harvest rates across season and fleets in the terminal year of stock assessment phase are used as 'current F' in the future projections. This caused too pessimistic results in this sample case, where end year F projections (fig. 7) showed over-exploitated condition, and multiplier F (Table 5) was small as 0.33 for achieving SPR 0.1. Harvest rates in the end year would have wide range according to the bootstrap results (fig. 5b), and might have positive bias as suggested retrospective analysis (fig. 8). Further discussion about reasonable definition of current F will be needed to avoid getting biased results in future projections.

In the previous stock assessment using tuned VPA (Yamada et al, 2006), estimated results in the last 3 years (2002-2004) were not used for future projections because estimation error is known to be accumulated toward recent years in the backward calculation used in tuned VPA. The future F pattern was determined by random draw or average from Fs during the previous 5-year period (1997-2001). Such procedures, as dropping the last a few years and averaging F in the next some recent years for current F, are common ways in the projections starting from the results by using VPA as well as MFCL. Because SS II uses population dynamics of forward calculation, its error accumulation pattern are different from that in VPA. Current F appropriate for SS II model and PBF data should be determined after enough observation of error structure of the results by SS II.

Technically, it is difficult to use arbitrary patterns of harvest rates as current F in the future projection by SSII. There is an option for setting arbitrary relative F pattern for future projection in SSII by setting 'relative F flag' to be 2 (ex. Table 4), but we observed problematic behavior of SS II in the setting, where estimated standard deviations were different from those in setting relative F flag to 1. We need further analysis or investigation of the option of 'relative F flag' in SS II. Usage of new option of 'F ballpark' in SS II newer than 2.00 might be another solution to avoid the too large harvest rates in the end year. When we set F ball park of end year of 2004 to 0.3, total exploitation rates in 2004 became smaller to 0.48. However, in setting F ball park as 0.2, calculation was not converged with large penalty, which suggested too much constraint in the last year F makes the model unstable.

Statistics and reference points to be focused

SS II can produce results on future projections as for spawning biomass, recruitments, depletion, total catch or total harvest rate and total catch per summary biomass with their estimated standard deviation (Methot 2007). However, we couldn't get exact probability distribution of the future statistics. Because some reference points such as Fssb are based on probability distribution of future focused statistics, SS II projections might not be able to deal with such a reference point. We can calculate nearly similar value even by SS II (eq. 3-4), but the probabilities was different from those produced from stochastic projections (Table 5).

Assumption of future recruitment

As SS II assumed log-normal distribution as historical recruitment deviation, it is straightforward to use the same assumption in the future projections. However, it is doubtful that higher or lower future recruitments than the historical highest or lowest level of recruitments can be really occurred in future because upper limit of recruitments might be determined biologically by possible environmental capacity, and hidden relationships between spawning biomass and recruitments might determine lower limit of recruitment. In fig. 1, both of upper 5% limits by SS II and stochastic projections were above the highest number of historical recruitments. According to the results of stochastic projections in fig. 1 (where σ_R of 1.1 was used as same as estimated in SS II), approximately 10% of recruitments were above the historical highest recruitment, and 2% was below the lowest. This suggests the possibility that extremely strong cohorts above the historically highest level, which will be occurred once per a decade approximately, might result in overestimation of future status of PBF stock.

What methods and results will be needed in the next assessment?

Based on the above discussion, we need to chose methods and results needed in the future projections for the next stock assessment for PBF. There should be many alternative methods and scenarios for future projections other than those presented in this document. It is preferable to use methods and scenarios under careful considerations of cost & benefits, and possible bias in those methods.

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Appendix A. Pseudo code for the future projections with R

 $N_{a,y}^{k}$ is the total number of future population at age *a* (=0, 0.25, 0.5, 0.75,...,19.75,20), year *y* (=2005,2005.25,2005.5,2005.75,2006,2006.25,2006.5,2006.75,2007,...,2043.75,2044) in the *k*th simulation (k=1,2, 3,...,500). Because SS II for the stock assessment for Pacific bluefin tuna is based on the time step with 4 seasons per year, future time series and age is considered as sequential vectors with 0.25 intervals. In the future time series, 2005, 2005.25, 2005.5 and 2005.75 are corresponding with 1st, 2nd, 3rd and 4th quarter in 2005, respectively. The time series before future projections is expressed by lower case and similar way of numbering. For example, numbers at age in the stock assessment phase can be expressed by $n_{0.1952}$, $n_{0.25.1952}$, $n_{0.5.1952}$.

For the first step of the stochastic future projections, all recruitments were determined on the basis of the assumption that recruitments occurs at the 1st quarter each year by random lognormal distribution with average of R_0 and standard deviation of σ^2 , which is estimated from SS2.

$$N_{0,y}^{k} = \begin{cases} \sim R_0 \exp(N(-\sigma^2/2, \sigma^2)) & \text{for } y = 2005, 2006, 2007, 2008, \dots, 2043, 2044 \\ 0 & \text{for } y = 2005.25, 2005.5, 2005.75, 2006.25, 2006.5, \dots, 2042.75 \end{cases}$$
(1)

In this calculation, we didn't consider any spawning-recruitment relationships because we can't find any relationships between estimated spawning biomass and recruitments in the current version of PBF assessment by SS2.

Given the estimates of number at age $n_{a,2005}$ except for recruitment of $N_{0,2005}$ by SS2 at the 1st quarter in 2005, expressed as, numbers at age in the first time step for future projections $(N_{a,2005,25}^k)$ can be calculated as the following.

$$N_{a,2005,25}^{k} = \begin{cases} N_{0,2005} \exp(-(F_{0,cur} + M_{0})) & \text{for } a = 0.25 \\ n_{a-0.25,2005} \exp(-(F_{a-0.25,cur} + M_{a-0.25})) & \text{for } a = 0.5, 0.75, \dots, 19.75 \\ n_{19,75,2005} \exp(-(F_{19,75,ur} + M_{19.75})) + n_{20,20055} \exp(-(F_{20,cur} + M_{20})) & \text{for } a = 20 \end{cases}$$
(2)

end for

for k=1, 2, 3, ..., 499, 500

$$N_{a,y}^{k} = \begin{cases} N_{a-0.25,2005.75}^{k} \exp(-(F_{a-0.25} + M_{a-0.25})) & \text{for } a = 0.25, 0.5, 0.75, \dots, 19.75 \\ N_{19.75,2005.75}^{k} \exp(-(F_{19.75} + M_{19.75})) + N_{20,2005.75}^{k} \exp(-(F_{20} + M_{20})) & \text{for } a = 20 \end{cases}$$
(3)

end for

end for

 F_a in the above algorism are depending on future fishing scenarios. We calculates F_a according to $f_{a,2004,s}$ calculated from estimated numbers at age $(n_{a,2004,s})$ and catch at age $(c_{a,2004,s})$ at *s*th quarter in 2004 by solving the following catch equation.

$$c_{a,2004,s} \cdot F_{multi} = \begin{cases} \frac{f_{a,2004,1}}{f_{a,2004,1} + M_a} (n_{a,2004,1} - n_{a-0.25,2003,4}) & \text{for s} = 1\\ \frac{f_{a,2004,s}}{f_{a,2004,s}} (n_{a,2004,s} - n_{a-0.25,2004,s-1}) & \text{for s} = 2,3,4 \end{cases}$$
(4)

 F_{multi} is a multiplier to the catch in the end year of 2004, which is derived from harvest multiplier estimated from SS II to archive target SSB level. The multipliers of F_{multi} should be 0 in no-fishing scenario and 1 in F_{2004} scenario. Future F_a is depending on the season as follows.

$$F_{a} = \begin{cases} f_{a,1} \text{ for } y = 2006, 2007, ..., 2047 \\ f_{a,2} \text{ for } y = 2006.25, 2007.25, ..., 2046.25 \\ f_{a,3} \text{ for } y = 2006.5, 2007.5, ..., 2046.5 \\ f_{a,4} \text{ for } y = 2006.75, 2007.75, ..., 2046.75 \end{cases}$$
(5)

After repeating the simulation 1000 times, future statistics of total biomass (B_y^k) and SSB (SSB_y^k) are calculated.

$$B_{y}^{k} = \sum_{a=0}^{20} w_{a} N_{a,y}^{k} \text{ for } y = 2006, 2006.25, ..., 2046.75, 2047 (6)$$

$$SSB_{y}^{k} = \sum_{a=0}^{20} w_{a} Q_{a} N_{a,y}^{k} \text{ for } y = 2006, 2006.25, ..., 2046.75, 2047 (7)$$

Then, an arithmetic average of $B_y^k \left(\sum_{k=1}^{1000} B_y^k\right)/1000$, and percentiles of 5, 50 and 95% are calculated.

Appendix table. Definition of Terms

а	a age index; a=0,0.25,0.75,1,1.25,,19.5,19.75,20+
у	y year index; y=1975, 1976,,2030,2031
k	k index for bootstrap replication number; k=0,,499,5003
$N_{a,v}^k$	Population numbers at the middle (?) of the quarter
$n_{a,y}^k$	Population numbers at the middle (?) of the quarter
M_a	M instantaneous rate of natural mortality (yr-1)
F_a	Instantaneous rate of fishing mortality in future
$F_{a,y}$	Instantaneous rate of fishing mortality in estimation phase
$C_{a,y}$	Estimated catch at age in weight in estimation phase
W_a	Weight at age of an individual at the middle (?) of the year
Q_a	Maturity rates at age
В	Stock biomass at the beginning of the year (?)
SSB	SSB spawning stock biomass at the beginning (?) of the spawning season

Appendix B. Control file of SS II applied to this simple data

```
# bluefin model with quarterly fisheries, annual cpue,
# Size data and catch data was updated at April, 2007
1
   # N growth patterns
   # N_sub-morphs per gender x growth pattern
1
1 #N areas
11111 11111 11111 #area for each fleet survey
# recruit design (G Pattern x birthseas x area) X (0/1 flag)
1 0 0 0 # season1-4
0 # recr dist interaction
  # domigration
0
0
   1 1
          #
              season1 # Movement Pattern by season x source x destination
           #
0
              season2
   1 1
           #
0
  1 1
              season3
0 1 1
           # season4
# Block parameters
    # Nblock patterns
1 1 2 2 # Blocks per pattern
#1981 2004
1994 2004 # vectors of beginning and ending years for blocks in design 1
1986 2004 # Fleet 7
1960 1980 1981 2004
1986 1993 1994 2004
# Entries that are common to all growth patterns, genders, and morphs
0.5 # Fraction of female to all growth pattern
    # submorph between/within stdev ratio
1
   # vector submorphdist
-1
# mortality and growth parms
    #last age for nat mortality young
0
3
    #first age for nat mortality old
0
     #age lmin
15 #age lmax
0.1 # SD add to LAA
1
  # CV pattern
#-4 #MG parm dev phase
   # Maturity option
1
3
   # First Mature Age
1
   # MGparam as offset
1
-1 # MGparams Dev Phase
# Read Mortality and Growth Parameters
# lo hi init prior pr type sd phase env-var use dev dvmnyr dvmxyr dvsddv block blktype
 0.01 2 1.6 1.6 0 0.8 - 3 0 0 0 0 0.5 0 0 # M young
 0.1 0.4 0.25 0.25 0 0.8 -3 0 0 0 0 0.5 0 0 # Natmort old
                              # Natural mortality for ages>=NMold
 3 200 22.5 26 0 1000 - 60
                                 0 0 0 0.5 0 0#Lmin
 201 400 257 257 0 1000 - 40
                                    0
                                       0 0 0.5 0 0 #Lmax
 0.01 0.65 0.1035 0.1035 0 0.2 -4 0
                                   0 0 0 0.5 0 0 #vbk
 0.01 0.25 0.08 0.08 0 999 -3 0 0 0 0.5 0 0 #CV for size age age, age<=AFIX
 0.01\ 0.25\ 0.08\ 0\ 0\ 0.8\ -3\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ FIX2
#len-wt and maturity
    -3 3
            0.00003 0.00003 0
                                0.8 -3
                                       0
                                            0
                                                0
                                                    0
                                                         0.5.0
                                                                0
                                                                     #wt len a
                                                          0.5 0
    -3 3
            2.9085 2.9085 0
                                                                     # wt len2 b
                                 0.8 - 3 0
                                             0
                                                 0
                                                     0
                                                                 0
                                                 0.5 0
    -3 3
            130 130 0
                                                        0
                                                                      #Maturity inflect
                        0.8 - 3
                               0
                                    0
                                        0
                                             0
            -.64
    -3 3
                            0
                                0.8 -3
                                       0
                                                0
                                                         0.5 0
                                                                0
                                                                     #Maturity 2
                    -.64
                                            0
                                                    0
    0
                      0
                                         0
                                             0
        1
             1
                 1
                          0.8 - 3 0
                                     0
                                                  0.50
                                                         0
                                                                       #egg/gram
    0
        1
             0
                 0
                     0
                          0.8 - 3
                                 0
                                     0
                                         0
                                              0
                                                  0.50
                                                          0
                                                                       #egg.gram slope
#pop*growth morph for the prop of each morph in each area
```

	-4	4	0	1	-1 9	9 - 3	0	0	0	0	0.5	0 0) # frac	tion mo	rph 1 to area 1
#po	p line	es fo	r the r	orop ass	signed t	o eac	h ai	rea							1
r.	-4	4	0	1	-1 9	9-3	0	0	0	0	0.5	0 0) # redis	tribution	hy area
	_4	4	Ő	1	-1 9	9_3	Ő	Ő	Õ	Õ	0.5	0 () # redis	tribution	by season
	_4	4	Ő	1	-1 9	9_3	0	0	0	Ő	0.5) # redis	tribution	by season
		- 1	0	1	1 0	0 3	0	0	0	0	0.5) $\#$ redis	tribution	by season
	-4	4	0	1	-1 9:	9-3 D 2	0	0	0	0	0.5		# redia	tribution	season
	-4	4	0	1	-1 9	9-3	0	0	0	0	0.5		μ μ μ μ μ	undution	_by_season
0	-4	. 4	0	1	-1 9	9-3	0	0	0	0	0.5	0 () # !		
0	# cus	t env	/ read												
0 #0	custor	n ble	ock re	ad											
#SF	l sect	ion													
1	#1=	beve	erton h	olt wit	h flat to	pp be	yon	d Bze	ero						
#	lo	hi	init	pric	r pr	type	sd	ph	ase						
	10	16	512	12	2 0	100	0	2	#]	Ln(R	.0)				
	0.2	1	0.99	0.7	0 1	000		3	#Steep	oness	5				
	0	2	0.6	0.	.6 0	1000		3	#sd r	ecru	itmen	ts			
	-5	5	0	0	0	1		-3	#env	link					
	-5	5	0	0	0	1		-4	#ini	t eq					
	-5	5	0	0	0	-1		-4	#fut	ure					
0	#SR	en	v link												
Ő	# SI	₹_en	v taro	ret											
1	# D	o re	cr de	, c t											
#	start	t_{rec}	end	v rec I ov	wer unr	or	nh	260							
#	105	$\frac{1}{2}$	200		20.20	2	рп	ase							
1.40	195	2 :	2004 V - C	+ -	20 20										
149	2 # F	irst_	Yr_fu	libias_	adj_in_	_MPI	J								
#1111	t_F														
#	Lo	Hı	ınıt	pr	or p	rtype	e s	d p	hase						
	0	.9	0.10	001 0.1	00000	0	1(0000	1 #	fishe	eryl c	atch			
	0	.2	0.00	0.000	0 0) 1	000	00	-1 #fi	sher	y2				
	0	.9	0.10	001 0.1	0001	0	10	0000	1 #	fishe	ery3	catch	l		
	0	.9	0.10	000 0.1	00000	0	10	0000	1 #	fishe	ery4	catch	l		
	0	.9	0.10	000 0.1	0000	0	10	0000	1 #	fishe	ery5	catch	l .		
	0	.9	0.10	000 0.1	0000	0	10	0000	1 #	fishe	ery6	catch	I I		
	0	.9	0.10	000 0.1	0000	0	10	0000	1 #	fishe	ery7	catch	l		
	0	.2	0.00	0.000	00 0) 1	000	00	-1 #fi	sherv	v8				
	0	.9	0.10	00.0.10	0000	0	10	0000	1 #	fishe	erv9 c	atch			
#Os	etun	.,	0.10	00 0.10	0000	Ũ				110114					
0	etup	0		0	0		1		30		#		fisherv1		ПТ
0		0		0	0		1		30		#		fishery?		Inel PS
0		0		0	0		1		20		#		fichory2		Jper 1 S Ituno DS
0		0		0	0		1		20		н 11		fishery 3		Julia PS
0		0		0	0		1		30		#		fishery4		J ITOII
0		0		0	0		1		30		#		fishery5		J PL
0		0		0	0		I		30		#		fishery6		J SetNet
0		0		0	0		1		30		#		fishery7		EPO
0		0		0	0		1		30		#		fishery8		TWLL
0		0		0	0		1		30		#		fishery9		Other
0		0		0	0		1		30		#		cpue1	JLL1	
0		0		0	0		1		30		#		cpue2	JLL2	
0		0		0	0		1		30		#		cpue3	J tuna I	PS
0		0		0	0		1		30		#		cpue4	J troll	
0		0		0	0		1		30		#		cpue5	EPO1	
Ő		0		0	õ		1		30		 #		cnue6	EPO?	
Ŭ		5		v	0		1		50		11'		epueo	11 02	
#201	ev or	nd ro	tentio	n											
#se	ovtr		dorata	u ntion d	lo mala		:	rror							
#se		0		Hflaat			m	1101							
	0	0	U A	#neet		D_1 P	c								
24	0	0	0	#Ilee	12 Jpn		3 0								
24	0	0	0	#flee	to NW	POP	3								
24	0	0	0	#tlee	et4 Jpn '	Iroll									

24	0	0	0	#fleet5 JpnPL
24	0	0	0	#fleet6 Set net
24	0	0	0	#fleet7 EPO
24	0	0	0	#fleet8
24	0	0	0	#fleet9
5	0	0	1	#fleet10 CPUE small JLL from 1994
5	0	0	1	#fleet11 CPUE dw JLL til 1994
5	0	0	4	#fleet12 CPUE Jpn troll
5	0	0	3	#fleet13 CPUE Jpn WPS
5	0	0	7	#Ilect14 CPUE EPO
Э	0	0	/	#Ilectis CPUE EPO
#20	مدماد	v		
#ag		0	0	#fleet1 II I
11	0	0	0	#fleet2 Inn Pel PS
11	0	0	0	#fleet3 NWPO PS
11	0	Ő	0	#fleet4 Inn Troll
11	0	Õ	0	#fleet5 Jnn PL
11	Ő	Õ	0	#fleet6 Set net
11	0	0	0	#fleet7 EPO
11	0	0	0	#fleet8
11	0	0	0	#fleet9
11	0	0	0	#fleet10 cpue JLL
11	0	0	0	#fleet11 cpue JLL
11	0	0	0	#fleet12
11	0	0	0	#fleet13
11	0	0	0	#fleet14
11	0	0	0	#fleet15
#LO) HI	I	NIT	PRIOR PR_type SD PHASE env-variable use_dev dev_minyr
dev	_max	yr	dev	v_stddev Block_Pattern
# JI	L log	gistic	cur	ve
19	2	50	200	
0.00	Л 8	50. 	30	
#IIC		ILL S		$\frac{0}{2}$ 0.000 2.000 0.000 0.000 0.000 0.000 0.0000 0.0000 0.0000 0.000000
#20	2	80 1	1/.	$5 \frac{1}{5} \frac{0}{999} \frac{2}{2} \frac{0}{0} $
#-0 # 1	-	+ 5	-4.2	78 - 4.5 - 999 - 2 - 0 - 0 - 0 - 0 - 0 - 0 - 0 - 0 - 0
#=1 # 1		2	7.0	7.8 0 999 2 0 0 0 0 0 0 0 $\frac{1}{10000000000000000000000000000000000$
#-1 #_5		, ,	-5	-5 0 999 2 0 0 0 0 0 0 0 μ ESC-wiDTT in(width)
#-5	í	, ,	9	9 0 999 -2 0 0 0 0 0 0 0 # FINAL set at the last size bin
11 3	-			
#fle	et 2.1	IK PS	S sel	ex option24
21.2	$\frac{1}{2}$	284.1		$60 \ 60 \ 0 \ 999 \ 2 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ \# \text{ peak}1$
-6	4		-6	-6 0 999 2 0 0 0 0 0 0 0 #Top
-1	9		5.3	5.3 0 999 2 0 0 0 0 0 0 0 #ASC-width ln(width)
-1	9		4.4	4.4 0 999 2 0 0 0 0 0 0 0 #DESC-WIDTH ln(width)
-5	9		-5	-5 0 999 2 0 0 0 0 0 0 0 # INITIAL sel at the first size bin
-5	9		-5	-5 0 999 2 0 0 0 0 0 0 0 # FINAL sel at the last size bin
#fle	et 3 1	NWP	PO P	S selex option24
21.2	2 2	284.1		125 125 0 999 2 0 0 0 0 4 2 # peak1
-6	4		1	1 0 999 2 0 0 0 0 4 2 #Top
-1	9		6.5	6.5 0 999 2 0 0 0 0 4 2 #ASC-width ln(width)
-1	9		9	9 0 999 2 0 0 0 0 4 2 #DESC-WIDTH ln(width)
-5	9		-5	-5 0 999 2 0 0 0 0 0 4 2 # INITIAL sel at the first size bin
-5	9		-5	-5 0 999 2 0 0 0 0 0 4 2 # FINAL sel at the last size bin
110		,	11	
i #fle	et 4 J	pn ti	roll s	elex option 24

block parameters
#-3 3 0 0 0 5 6 #
20 280 124.5 124.5 0 999 2 # peak1 Fleet3 1986-1993
20 280 185.8 185.8 0 999 2 # peak1 Fleet3 1994-2004
-6 4 -4.8 -4.8 0 999 4 #Top
-6 4 -6 -6 0 999 2 #Top
-1 9 7.2 7.2 0 999 2 #ASC-width ln(width)
-1 9 8.1 8.1 0 999 2 #ASC-width ln(width)
-1 9 8.3 8.3 0 999 2 #DESC-WIDTH ln(width)
-1 9 7 7 0 999 2 #DESC-WIDTH ln(width)
-5 1 -5 -5 0 999 2 # INITIAL selectivity at the first size bin
-5 1 -5 -5 0 999 2 # INITIAL selectivity at the first size bin
-5 1 -5 -5 0 999 2 # FINAL selectivity at the last size bin
-5 1 -5 -5 0 999 2 # FINAL selectivity at the last size bin
#20 60 40 35.7 0 999 2 # peak1 Fleet5 1994-2004
#-22 10 -5 -6 0 999 2 #Top
#2 4 2.5 3 0 999 2 #ASC-width ln(width)
#2 4 2.5 3 0 999 2 #DESC-WIDTH ln(width)
#-5 1 -5 -5 0 999 2 $#$ INITIAL selectivity at the first size bin
#-5 1 -5 -5 0 999 -2 # FINAL selectivity at the last size bin
20 280 88.8 88.8 0 999 2 # peak1 Fleet7 1985-2004
-6 4 -0.2 -0.2 0 999 2 #Top
-1 9 5.8 5.8 0 999 2 #ASC-width ln(width)
-1 9 6.3 6.3 0 999 2 #DESC-WIDTH In(width)
-5 1 -5 -5 0 999 2 #INITIAL selectivity at the first size bin
-5 I -5 -5 0 999 2 # FINAL selectivity at the last size bin
<pre>#0 #///////////////////////////////////</pre>
00000 00000 00000
row 4 multiplier for lencomp effective N (set to 1.0 for no effect)
11111 11111 11111
row 5 multiplier for agecomp effective N (set to 1.0 for no effect)
00000 00000 00000
row 6 multiplier for size-at-age effective N (set to 1.0 for no effect)
00000 00000 00000
#
30 # DF for T dist used to scale discard deviations
30 # DF for T dist used to scale mean body wt dev
1 #max lambda phases
0 # include (1) or not (0) the constant offset for Logs(s) in the Log(like) calculation
survey lambdas
discard lambdas
0 # mean body wt
#lentreq lambda
#age treq lambda
mean size at age

00000 00000 00000 1 #init equlib F 1 #rec lambda 0 #parm prior lambda 0 #prior dev timeseries lambda 100 #crashpen lambda 0.9 #max F 999

	starter.ss2											
	(abbr.)											
22	1 #Forecast option (moved from forecast.ss2 and change index											
	#assignment)											
	#0 = no forecast											
	#1 = use F(spr) for forecast											
	#2 = use F(msy) for forecast											
	#3 = use F(btarget) for forecast											
	#4 = use ending year F for forecast											
23	1 #MSY option (moved from forecast.ss2 and change index assignment)											
	#0 = no MSY calc											
	#1 = set F(msy) = F(spr)											
	#2 = calc F(msy)											
	#3 = set F(msy) = F(Btarget)											
	#4 = set F(msy) = ending year F											
24	1 #Do output for west coast groundfish rebuilder package											
	#0=skip, but items two below still must exist to be read. Also, if											
	#forecast is turned off, then this should be turned off also.											
	#1=do											
25	-1 # year declared for rebuilder package											
	#This is the first year for which catch could have been set to zero											
	#(Ydecl). SS2 will output the age composition at the beginning of this											
	#year.											
	#Default = -1 will set to 1999											
26	-1 #start year for rebuilder package											
	#This is the year from which the rebuilder package will start											
	#simulations (Yinit). SS2 will output the age composition at the											
	#beginning of this year.											
	#) etails = -1 will set to endur+1											
	Provent and											
1	forecast.ss2											
1	forecast.ss2 0.1 # Target SPR 40 # Number of forecast upon											
1 2 2	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years											
1 2 3	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less # then or equal to total forecast years											
1 2 3	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below.											
1 2 3 4 5	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1											
1 2 3 4 5 6	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Eraction log-bias adjustment After End Year											
1 2 3 4 5 6 7	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less 40 # Number of forecast years with stddev, must be less 41 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option											
1 2 3 4 5 6 7 8	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less 40 # Number of forecast years with stddev, must be less 41 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 10 # Bottom of 400 Option											
1 2 3 4 5 6 7 8 9	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less 40 # Number of forecast years with stddev, must be less 41 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 10 # Bottom of 400 Option 1 # OY Scalar to ABC											
1 2 3 4 5 6 7 8 9 10	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less 40 # Number of forecast years with stddev, must be less 41 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 10 # Bottom of 400 Option 1 # OY Scalar to ABC -3 # First year (relative to endyr) for averaging fishery selectivity											
1 2 3 4 5 6 7 8 9 10 11	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less 40 # Number of forecast years with stddev, must be less 41 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 10 # Bottom of 400 Option 1 # OY Scalar to ABC -3 # First year (relative to endyr) for averaging fishery selectivity 0 # last year for average selectivity											
1 2 3 4 5 6 7 8 9 10 11 12	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less 40 # Number of forecast years with stddev, must be less 41 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 10 # Bottom of 400 Option 1 # OY Scalar to ABC -3 # First year (relative to endyr) for averaging fishery selectivity 0 # last year for average selectivity 1 # Relative F Flag											
1 2 3 4 5 6 7 8 9 10 11 12	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less 40 # Number of forecast years with stddev, must be less 41 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 10 # Bottom of 400 Option 1 # OY Scalar to ABC -3 # First year (relative to endyr) for averaging fishery selectivity 0 # last year for average selectivity 1 # Relative F Flag # 1=set relative F from endyr											
1 2 3 4 5 6 7 8 9 10 11 12	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 1 # OY Scalar to ABC -3 # First year (relative to endyr) for averaging fishery selectivity 0 # last year for average selectivity 1 # Relative F Flag # 1=set relative F from endyr # 2= use relative F read below											
1 2 3 4 5 6 7 8 9 10 11 12	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 10 # Bottom of 400 Option 10 # Bottom of 400 Option 10 # Bottom of 400 Option 10 # Isst year (relative to endyr) for averaging fishery selectivity 0 # Last year for average selectivity 1 # Relative F Flag # 1=set relative F from endyr # 2= use relative F read below # then enter the pattern of relative harvest rate to be used when Option 2											
1 2 3 4 5 6 7 8 9 10 11 12	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 10 # Bottom of 400 Option 10 # Bottom of 400 Option 1 # Cory Scalar to ABC -3 # First year (relative to endyr) for averaging fishery selectivity 0 # last year for average selectivity 1 # Relative F Flag # 1=set relative F from endyr # 2= use relative F read below # then enter the pattern of relative harvest rate to be used when Option 2 # above is selected. Otherwise, the pattern of harvest rates across seasons and fleets in the endyear											
1 2 3 4 5 6 7 8 9 10 11 12	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Top of 400 Option 10 # Bottom of 400 Option 10 # Bottom of 400 Option 1 # OY Scalar to ABC -3 # First year (relative to endyr) for averaging fishery selectivity 0 # last year for average selectivity 1 # Relative F Flag # 1=set relative F from endyr # 2= use relative F read below # then enter the pattern of relative harvest rate to be used when Option 2 # above is selected. Otherwise, the pattern of harvest rates across seasons and fleets in the endyear # is used when finding Fmsy, Fspr and doing the forecast.											
1 2 3 4 5 6 7 8 9 10 11 12 13	 forecast.ss2 0.1 # Target SPR 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 1 # OY Scalar to ABC -3 # First year (relative to endyr) for averaging fishery selectivity 0 # last year for average selectivity 1 # Relative F Flag # 1=set relative F from endyr # 2= use relative F read below # then enter the pattern of relative harvest rate to be used when Option 2 # above is selected. Otherwise, the pattern of harvest rates across seasons and fleets in the endyear # is used when finding Fmsy, Fspr and doing the forecast. 0 0 0 0 0 0 0 0 0 0 											
1 2 3 4 5 6 7 8 9 10 11 12 13	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below 1 # fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 10 # Bottom of 400 Option 10 # Bottom of 400 Option 10 # Bottom of ABC -3 # First year (relative to endyr) for averaging fishery selectivity 0 # last year for average selectivity 1 # Relative F Flag # 1=set relative F from endyr # 2= use relative F read below # then enter the pattern of relative harvest rate to be used when Option 2 # above is selected. Otherwise, the pattern of harvest rates across seasons and fleets in the endyear # is used when finding Fmsy, Fspr and doing the forecast. 0 0 0 0 0 0 0 0 # relative harvest rate for fleet0-9 in season 1 0 0 0 0 0 0 0 0 # relative harvest rate for fleet0-9 in season 2											
1 2 3 4 5 6 7 8 9 10 11 12 13	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below 1 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 10 # Bottom of 400 Option 11 # OY Scalar to ABC -3 # First year (relative to endyr) for averaging fishery selectivity 0 # last year for average selectivity 1 # Relative F Flag # 1=set relative F from endyr # 2= use relative F read below # then enter the pattern of relative harvest rate to be used when Option 2 # above is selected. Otherwise, the pattern of harvest rates across seasons and fleets in the endyear # is used when finding Fmsy, Fspr and doing the forecast. 000000000 <t< th=""></t<>											
1 2 3 4 5 6 7 8 9 10 11 12 13	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below 1 # recruitment deviation emphasis see below 1 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 10 # Bottom of 400 Option 10 # Bottom of 400 Option 10 # Bottom of 400 Option 11 # OY Scalar to ABC -3 # First year (relative to endyr) for averaging fishery selectivity 0 # last year for average selectivity 1 # Relative F Flag # 1=set relative F read below # then enter the pattern of relative harvest rate to be used when Option 2 # above is selected. Otherwise, the pattern of harvest rates across seasons and fleets in the endyear # is used when finding Fmsy, Fspr and doing the forecast. 000000000 # relative harvest rate for fleet0-9 in s											
1 2 3 4 5 6 7 8 9 10 11 12 13	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below 1 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 10 # Bottom of 400 Option 10 # Bottom of 400 Option 1 # OY Scalar to ABC -3 # First year (relative to endyr) for averaging fishery selectivity 0 # laset for average selectivity 1 # Relative F Flag # 1=set relative F from endyr # 2= use relative F read below # then enter the pattern of relative harvest rate to be used when Option 2 # above is selected. Otherwise, the pattern of harvest rates across seasons and fleets in the endyear # is used when finding Fmsy, Fspr and doing the forecast. 0 0 0 0 0 0 0 0 # relative harvest rate for fleet0-9 in season 1 0 0 0 0 0 0 0 0 # relative harvest rate for fleet0-9 in season 2 0 0 0 0 0 0 0 0 <t< th=""></t<>											
1 2 3 4 5 6 7 8 9 10 11 12 13 14	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Top of 400 Option 10 # Bottom of 400 Option 10 # Bottom of 400 Option 1 # OY Scalar to ABC -3 # First year (relative to endyr) for averaging fishery selectivity 0 # last year for average selectivity 1 # Relative F Flag # 1=set relative F read below # then enter the pattern of relative harvest rate to be used when Option 2 # above is selected. Otherwise, the pattern of harvest rates											
1 2 3 4 5 6 7 8 9 10 11 12 13 13	forecast.ss2 0.1 # Target SPR 40 # Number of forecast years 40 # Number of forecast years with stddev, must be less # than or equal to total forecast years. Nearly all model dimensioning uses this value 1 # recruitment deviation emphasis see below 1 # recruitment deviation emphasis see below 1 # Fraction log-bias adjustment Before End Year+1 0 # Fraction log-bias adjustment After End Year 0 # Top of 400 Option 10 # Bottom of 400 Option 1 # OY Scalar to ABC -3 # First year (relative to endyr) for averaging fishery selectivity 0 # last year for average selectivity 1 # Relative F Flag # 1=set relative F read below # then enter the pattern of relative harvest rate to be used when Option 2 # above is selected. Otherwise, the pattern of harvest rates across seasons and fleets in the endyear # is used when finding Fmsy, Fspr and doing the forecast. 000000000 # relative harvest rate for fleet0-9 in season 1 000000000 # relative harvest rate for fleet0-9 in season 2 000000000 # relative harvest rate for fleet1-9 in season 4											
1 2 3 4 5 6 7 8 9 10 11 12 13 13	Operating 1 with set to ensyle 1 A with set to ensyle 1											

Table 1. Example files of starter.ss2 and forecast.ss2 when assuming SPR0.1.

-1 -1 -1 -1 -1 -1 -1 -1 -1	#	year 2 seaso	n 1	fleet	1-9
(abbr.)					

Table 2. Example files of starter.ss2 and forecast.ss2 when assuming no-fishing for projections. Parts expressed by (abbr.) are same as the settings shown in Table 1.

	starter.ss2
	(abbr.)
22	4 #Forecast option (moved from forecast.ss2 and change index
	(abbr.)

	forecast.ss2	
	(abbr.)	
12	2 # Relative F F	lag
13	0000000000	# relative harvest rate for fleet0-9 in season 1
	0000000000	# relative harvest rate for fleet0-9 in season 2
	0000000000	# relative harvest rate for fleet0-9 in season 3
	0000000000	# relative harvest rate for fleet1-9 in season 4
14	999	
	(abbr.)	

Table 3. Example files of starter.ss2 and forecast.ss2 when assuming F 2004. Parts expressed by (abbr.) are same as Table 1.

	starter.ss2
22	(abbr.) 4 #Forecast option (moved from forecast ss? and change index
22	(abbr.)

	forecast.ss2	
	(abbr.)	
12	1 # Relative F	Flag
13	0000000000	# Average harvest rate for fleet 0-9 in season 1 from 2002 to 2004
	0000000000	# Average harvest rate for fleet 0-9 in season 1 from 2002 to 2004
	0000000000	# Average harvest rate for fleet 0-9 in season 1 from 2002 to 2004
	0000000000	# Average harvest rate for fleet 0-9 in season 1 from 2002 to 2004
14	999	
	(abbr.)	

Table 4. Example files of starter.ss2 and forecast.ss2 when assuming current F from 2002 to 2004. Parts expressed by (abbr.) are same as the settings shown in Table 1.

ſ		starter.ss2
		(abbr.)
	22	4 #Forecast option (moved from forecast.ss2 and change index
		(abbr.)

	forecas	t.ss2											
	(abbr.)												
12	2 2 # Relative F Flag												
13	0.01	0.09	0.07	0.02	0.03	0.02	0.10	0.00	0.01				
	0.01	0.54	0.00	0.46	0.11	0.01	0.00	0.00	0.01				
	0.01	0.17	0.00	0.19	0.03	0.00	0.00	0.00	0.00				
	0.07	0.41	0.03	0.16	0.10	0.03	0.05	0.26	0.00				
	# The above values are average exploitation in 2002-2004												
14	999												
	(a	bbr.)											

Table 5. Probability that the SSB will fall below the histrorical lowest SSB during one or more years of the projection period. F multiplier means a scalar multiplied by matrix of exploiatation rates by fleet and season to the last year of 2004, for which target SSB level will be archieved. In the scenarios of blank columns, calculation was not conducted because of limitation of time and computer resources. In the attached figure, x axis was shown by F multiplier to the F at the last year 2004, for achieving future target SSB level.

	F	SS II	Stochastic	Stochastic with	Stochastic with resampled
	multiplier			bootstrap	recruitment
No fishing	0	0.06	0.00	0.00	0.00
SPR 0.1	0.33	0.07	0.00	0.00	0.00
SPR 0.05	0.44	0.12	0.01	0.04	0.00
SPR 0.03	0.51	0.18	0.31	0.47	0.24
SPR 0.01	0.66	0.60	1.00	0.64	1.00
Current F	1.00	1.00	1.00	1.00	1.00
Pr[SSB ₂₀₄₀ <ssb<sub>observed F]</ssb<sub>					
	F	SS II	Stochastic	Stochastic with	Stochastic with resampled
	multiplier			bootstrap	recruitment
No fishing	0	0.00	0.00	0.00	0.00
SPR 0.1	0.33	0.01	0.00	0.00	0.00
SPR 0.05	0.44	0.03	0.00	0.00	0.00
SPR0.03	0.51	0.09	0.01	0.02	0.01
SPR 0.01	0.66	0.54	0.59	0.99	0.85
Current F	1.00	1.00	1.00	1.00	1.00

Pr[SSB_{future}<SSB_{observed}|F]



Fig. 1. Comparison of the assumed future recruitments between SS2 and stochastic projection. (a) Future trajectories of SSB. Green sold and dotted lines show median and 90% confidence interval by by stochastic projection. Blue line shows arthmetric average of SSB estimated from stochastic projection. Estimated average and $\pm 1.64\sigma$ from SS2 are shown by black and gray dotted lines, respectively. (b) Probability distribution of SSB in 2040. black: SS2, green: stochastic projection



Fig. 2. Comparison between projected numbers at age by SS II and stochastic projections. Gray and balck dotted lines show simple average and median of stochastic projections, respectively, and black solid lines show results by SS2.



Fig. 3. Comparison between projected SSBs and probability distributions in 2040 by SS II and stochastic projection under the scenario of no-fishing. Figure legend is same as in fig. 1.



Fig. 4. Comparison between projected SSBs and probability distributions in 2040 by SS II and



stochastic projection under the scenario of SPR=0.1. Figure legend is same as in fig. 1.

Fig. 5. Total exploitation rates estimated by bootstrap 150 times tirals. (a) Historical exploitation rates (b) total exploitation rates in the end year of 2004. Average and point estimate of total exploitation rate in 2004 was also shown by two arrows.



Fig. 6. Results by stochastic projection with bootstrap estimates. (a) trajectories of SSB. (b) Probability distribution of SSB in 2011



Fig. 7. Comparison between projected SSBs and probability distributions in 2040 by SS II and stochastic projection under the scenario of F in the year. Figure legend is same as in fig. 1.



Fig. 8. Retrospective pattern of the total exploitation rates. Results of exploitation rates in minus 2 years data was not shown because of no convergence of the model.



Fig. 9. Comparison of stochastic projections between different recruitment scenarios of lognormal (gray) and resampling (green) under the scenario of SPR0.1. Trejectories of future recruitment (a) and SSB (b) are shown.