# Using information for 'data-rich' species to inform assessments of 'data-poor' species through Bayesian stock assessment methods 

André E Punt, David C Smith and Matthew T Koopman

Department of
Primary Industries

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# Using information for 'data-rich' species to inform assessments of 'data-poor' species through Bayesian stock assessment methods 

Andre E Punt ${ }^{1}$, David C Smith ${ }^{2}$ and Matthew T Koopman ${ }^{2}$

## Address

${ }^{1}$ School of Aquatic and Fishery Sciences
Box 355020
University of Washington
Seattle, WA 98195-5020
USA

[^0]Queenscliff, VIC 3225
Australia

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## Non Technical Summary

## 2002 / 094: Using information for 'data-rich' species to inform assessments of 'data-poor' species

 through Bayesian stock assessment methodsPrincipal Investigator: Dr David C Smith<br>Address: Primary Industries Research Victoria<br>PO Box 114<br>Queenscliff, VIC 3225<br>Australia<br>Tel: 61-3-5258-0266 Fax 61-3-5258-0270<br>Email: David.c.Smith@dpi.vic.gov.au<br>Co-Investigator:<br>Assoc. Prof. Andre Punt<br>Address: $\quad$ School of Aquatic and Fishery Sciences<br>Box 355020<br>University of Washington<br>Seattle, WA 98195-5020<br>USA<br>Tel: 1-206-221-6319 Fax: 1-206-685-7471<br>Email: aepunt@u.washington.edu

## Objectives:

1. Expand the database of production parameters for SEF-like species.
2. Develop prior probability distributions for steepness and the coefficient of variation about the stockrecruitment relationship using Bayesian meta-analysis.
3. Develop a Bayesian framework within which the results for data-rich species can 'inform' assessments for data-poor species.
4. Apply the framework to three case-studies to determine the robustness of the framework.
5. Test the framework by means of Monte Carlo simulation

## Non Technical Summary:

The 'Integrated Analysis' approach to fisheries stock assessment has become the most common method for conducting stock assessments in the Southern and Eastern Scalefish and Shark Fishery (SESSF) for several reasons. These reasons include that 'Integrated Analysis' has been shown using Monte Carlo simulation to perform better than alternative methods such as production models. 'Integrated Analysis' can make use of most of the data available for assessment purposes. However, predictions based on this approach will be very uncertain unless biological information and an extensive time-series of catch-at-age data and abundance indices are available. For many of the 'low priority' species in the SESSF (and the by-product and by-catch species), the only data collected 'routinely' are catch and effort statistics, in most
cases length-frequencies (at least for quota species), and in some cases 'snapshots' of age-composition. It is, and will likely remain, almost impossible to conduct stock assessments for all of the species for which they are needed unless inferences for 'data-poor' stocks are based in part on information for 'data-rich' stocks, and, more generally, on knowledge for other species and stocks. Formally, this information can be included in assessments in the form of 'prior distributions' (or penalty functions).
This report considers two ways to include 'prior distributions' in assessments. First, whether the data for 'data-rich' species/stocks can be used to develop prior distributions for some of the key input parameters included in stock assessment models is examined. The results indicate that the best way to estimate the rate of natural mortality is Hoenig's equation (or the variant thereof based on the data analysed in this report) while the method developed by Pauly seems very sensitive to whether data for species with high $M$ are included in the analysis. Several relationships among biological parameters are identified, but the predictive ability of most of these is reasonably weak. The fact that $\ell_{\infty}$ was found to correlate well with several other biological parameters highlights the importance of conducting ageing studies for as many species as possible (even if the sample sizes are only sufficient to estimate the parameters of a growth curve) and/or to conduct a catch-curve analysis.
Stock and recruitment data were analysed to estimate priors for the steepness of the stock-recruitment relationship. These data suggest that steepness for Clupeiformes, Pleuronectiformes, and Gadiformes is higher than that for "other" species. This needs to be accounted for when selecting the data on which to base priors for steepness for SESSF species as only blue grenadier is in this group of "high steepness" fishes. In contrast, steepness is notably lower for species not in these three families and this should therefore be expected of most of the species in the SESSF. The reasons for this difference are unclear. Also no covariates that explain steepness were found (although the sample sizes for some of the covariates are quite small) except whether the fish is a Clupeiform, Pleuronectiform, or Gadiformes. The results can be used to develop base-case priors for steepness and the extent of variation in recruitment, and the values for these quantities to be used in tests of sensitivity. It is recommended that if a single default point estimate for steepness is to be used in a stock assessment based on the Beverton-Holt stockrecruitment relationship, that default should be 0.907 for Clupeiformes, Gadiformes and Pleuronectiformes and 0.757 for 'other' species.
The second approach to make use of information for 'data-rich' species when conducting assessments for 'data-poor' species is to conduct assessments of several species ('data-rich' and 'data-poor') simultaneously and to impose penalties on the differences in biological parameters among species/stocks.

It is clearly not possible to share values of parameters (such as the age-at-maturity) among different species. However, it seems plausible that if multiple stocks/species are exploited by the same fleet (where fleet is defined as a group of vessels fishing in the essentially the same fishing grounds at the same time), the trend in fishing mortality for that fleet should be similar for all of the stocks/species. It also seems likely that the annual deviations in recruitment about the stock-recruitment relationship for different stocks/species would be correlated (positively or negatively) due to the impact of common environmental variables, and that selectivity as a function of length (before discarding) should be relatively similar across various fleets. However, there are also good reasons related to the behaviour of fishers and fish, that parameters for different stocks/species should not be identical. The approach of this report therefore allows for stock-specific values for all of the model parameters, but adds penalties on how different the values for these parameters may be across stocks. This basic approach has been used previously for multiple stocks of the same species, but this is the first time multiple stocks of several species have been assessed simultaneously.
The technical details of the method are described, tested by means of simulation and applied to data for eight stocks (seven species) in the SESSF. The example application is based on five fleets (four trawl fleets and a non-trawl fleet) and eight stocks: blue grenadier (data-rich), eastern gemfish, spotted warehou and pink ling (data-moderate), and western gemfish, mirror dory, king dory and ocean perch (data-poor). The data available included catch and effort, discard rates, length-frequencies and age-compositions by fleet and stock, values for biological parameters, and survey estimates of the spawning biomass of blue grenadier in 1994 and 1995.
The results of the simulations and the example application confirm that there is value in imposing crossstock/species constraints. However, the effects may be quite small, and the consequences, in terms of bias and precision, may be negative for some of the 'data-rich' species. Nevertheless, the expectations of

Information for 'data-rich' species to inform assessments of 'data-poor' species
improved stability and precision are largely borne out - the results of the analyses, specifically the trends in the abundance of ocean perch, king dory and western gemfish, differ fairly substantially when amongstock constraints are taken into account. Furthermore, the precision of some of the key outputs from an assessment (the trends in spawning biomass and spawning biomass expressed relative to the unfished level) for the 'data-poor' stocks is much greater when among-stock constraints are taken into account
The approaches outlined in this report cannot turn a 'data-poor' stock into a 'data-rich' stock, but will, at the very least, be of value to identify species that should be the focus of increased data collection and analysis. In this study, the results suggest that the stock of western gemfish may be depleted (although it should be noted that the fishery is at the eastern end of the species' range) while there are declining trends in ocean perch and king dory that warrant further data collection and analysis.

## Outcomes Achieved

The main outcome from this project is the development and testing of an analytical tool that can be applied to the many low value species in the SESSF to help formally assess the status of the stocks. As SESSRAG is currently developing harvest strategies for the fishery, the results of this project provide an improved ability to advise fishery managers on appropriate indicators and reference points for some "data-poor" quota species that will probably provide better indicators of trend that either CPUE or agestructure alone, because the approach integrates all of the available data. Ultimately it will provide a costs-effective means of undertaking more formal assessments of the many SESSF species that currently have no formal assessment. At the very least, the approach developed here will be of value to identify species that should be the focus of increased data collection and analysis.
Benefits of this project also flow to many of the fisheries managed by AFMA and state agencies because the priors and relationships developed can be used for assessments other than for those of SESSF species.

## Keywords:

Bayesian, Hierarchical meta-analysis. Multi-species modelling, Simulation evaluation, Technical interactions

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## Chapter 1: Introduction

## Background

The 'Integrated Analysis' approach to fisheries stock assessment has become the most common method for conducting stock assessments in the SESSF for several reasons (Smith et al. 2001). These reasons include that 'Integrated Analysis' has been shown using Monte Carlo simulation to perform better than alternative methods such as production models (Punt et al. 2002a). 'Integrated Analysis' can make use of most of the data available for assessment purposes. However, predictions based on this approach will be very uncertain unless biological information and an extensive time-series of catch-at-age data and abundance indices are available. Furthermore, there is a need for 'contrast' in the data to enable biomass to be estimated reliably. For many of the 'low priority' species in the SESSF (and the by-product and bycatch species), the only data collected 'routinely' are catch and effort statistics, in most cases lengthfrequencies (at least for quota species), and in some cases 'snapshots' of age-composition. Application of 'Integrated Analysis' to the data for these species will therefore lead to highly uncertain results, even though some level of ongoing assessment is required to meet strategic assessment under EPBC.
The question is, of course, why we should aim for a formal stock assessment of these species. Formal stock assessments provide crucial insights into the status of fish stocks, but, more importantly, provide a predictive capacity that enables managers to assess alternate management or harvest strategies. The results of population projections form the key basis for management advice. Simply put, stock assessment allows managers to be proactive rather than reactive.
Koopman et al. (2000) demonstrated the utility of using information for 'similar' species when conducting assessments for SESSF species. The key parameters that form the basis for assessments of fish stocks are the virgin biomass, the rate of natural mortality, and the 'steepness' of the stock-relationship relationship. Of these parameters, 'steepness' is the hardest to estimate, usually due to lack of data 'contrast'.

However, information for other stocks / species can be used to infer ranges for steepness (and the extent of variability in recruitment). These ranges can be used in the assessments of SESSF species so long as the other species are 'similar' to the SESSF species for which information on steepness is needed. Information on natural mortality, steepness, the extent of recruitment variability, and a range of other biological parameters was obtained for 15 of the SESSF quota species and 52 other stocks / species by Koopman et al. (2000). A simple formula was developed for identifying 'similar' stocks / species to a species for which production parameters are required, using information on maximum age, depth, habitat preference, and diet. The results of the application of this algorithm were consistent with subjective judgements. In addition, an algorithm was developed for constructing prior probability distributions for steepness and the extent of variability in recruitment using the results from analyses for similar species. The resultant distributions from this algorithm could be used in Bayesian stock assessments and as the basis for sensitivity tests when applying other methods of stock assessments.

Koopman et al (2000) argued that the approach used to identify similar species appeared to be relatively successful with few obvious 'errors'. However, the approach used to develop probability distributions could be extended further in several key ways: (a) the database developed during the project could be extended considerably to include more stocks, and (b) the approach for developing probability distributions only took account of whether a species was similar to the SESSF species under consideration. It ignored whether the estimates of steepness for the 'other' species were sufficiently reliable themselves to be used to infer values for SESSF species. For example, several of the values used were themselves 'guestimates' rather than estimates based on actual data. Koopman et al. (2000) noted that it should be possible to apply Bayesian hierarchical meta-analysis (Gelman et al., 1995; Liermann and Hilborn, 1997) to obtain prior probability distributions for steepness and recruitment variation.
Development of prior probability distributions for use in Bayesian stock assessment approaches (e.g. Punt and Hilborn, 1997; McAllister and Ianelli, 1997) should reduce uncertainty when conducting assessments. For example, prior probability distributions for steepness and the extent of depensation are already included in the assessment of the eastern stock of gemfish (Smith and Punt, 1998). However,
given the lack of data (other than catches) for many of the species in the SESS, even having prior probability distributions for production parameters will be insufficient to restrict the value of the virgin biomass, $\boldsymbol{b}_{0}$. For such species, uncertainty will remain high if assessments are only based on the data for the species for which an assessment is required.

In principle, the assessments for well-studied ('data-rich') species (e.g. blue grenadier, flathead, and eastern gemfish in the SESSF) provide information on, for example, trends in fishing intensity by area and even perhaps the relative strength of recruitment that are relevant to assessments of 'data-poor' species. However, this information cannot presently be included formally in stock assessments for 'data-poor' species.

## Need

Over 300 species are caught in the SESSF, of which around 100 have commercial value. Twenty five species comprise around $90 \%$ of the landed catch. Each year, however, quotas are set for only around 17 species. Formal stock assessments (that may not occur every year) are available for ten of these species. No formal assessment is undertaken for any of the remaining quota species and for some of the more important non-quota species, and the only assessment for these species is an investigation of trends in catch, effort, and size distribution, and anecdotal input from scientists and industry. There are insufficient resources to undertake formal stock assessments for the wide range of commercial species landed in the SESSF. Yet, each of these species is an important component of the catch of fishers. If the fishery is to continue to operate in its current form, and meet the strategic assessments required under the EPBC Act, some form of formal assessment is required.

Koopman et al. (2000) demonstrated the utility of using information for 'similar' species when conducting assessments for SESSF species. Using key parameters such as the virgin biomass, the rate of natural mortality, and the 'steepness' of the stock-relationship relationship for a wide range of species, and a simple formula for identifying 'similar' stocks / species, an algorithm was developed for constructing prior probability distributions for these parameters. The resultant distributions could be used in Bayesian stock assessments and as the basis for sensitivity tests when applying other methods of stock assessments. The current project refines the prior distributions for two of the production parameters and develops and test methods of stock assessment that use the results of assessments for well-studied species in a formal manner to inform assessments of 'data-poor' species.

## Objectives

- Expand the database of production parameters for SEF-like species.
- Develop prior probability distributions for steepness and the coefficient of variation about the stockrecruitment relationship using Bayesian meta-analysis.
- Develop a Bayesian framework within which the results for data-rich species can 'inform' assessments for data-poor species.
- Apply the framework to three case-studies to determine the robustness of the framework.
- Test the framework by means on Monte Carlo simulation


## Organisation of this report

The body of this report is organised into a series of discrete chapters.
Chapter 2 provides a meta-analysis of production parameters and presents an examination of biological parameters and their inter-relationships. In the absence of comprehensive data on the population dynamics of a stock that would allow the application of formal stock assessment methods, 'rapid assessments' are an increasingly important tool that rely on basic life-history parameters only. Commonly-used approaches include the empirical formulae developed by Pauly (1980) and Hoenig (1983) to calculate $M$.

Chapter 3 conducts a Bayesian hierarchical meta-analysis of stock and recruitment data to develop prior distributions for steepness and the extent of variation in recruitment using data for other stocks /species. Hierarchical meta-analysis is a Bayesian technique that can be used to combine data from several
independent sources (species / stocks) and represent the outcome in the form of a probability distribution for a quantity of interest.
Chapter 4 summarizes the data that will be used in the example application of the method developed in Chapter 5. SESSF data, including catches, discards, age- and length-compositions and biological information are extracted for blue grenadier, eastern and western gemfish, pink ling, spotted warehou, mirror dory, king dory, and offshore ocean perch. Data are presented separately for four trawl fleets and one non-trawl fleet.

Chapter 5 describes an approach that informs the assessment of 'data-poor' species using information for 'data-rich' species. It therefore extends previous work based on several stocks of the same species to several species. An example application of the method based on the data summarized in Chapter 4 is included in this chapter.

It is necessary to evaluate all new methods of analysis to show that they perform at least as well as, and preferably better than, existing methods, given that a poor analysis method can have substantial economic, social, and biological consequences. Chapter 6 conducts this evaluation for the method outlined in Chapter 5.

## Chapter 2: Meta-Analysis of Production Parameters

## Introduction

Biological parameters are often used to characterise species and stocks. The most common of these parameters are: a) the parameters of the growth curve (e.g. $L_{\infty}, \kappa$ and $t_{0}$ for the von Bertalanffy form of the relationship between length and age), b) the rate of natural mortality, $M, \mathrm{c}$ ) the age- and length-atmaturity, $T_{\mathrm{m}}$ and $L_{\mathrm{m}}$, and d) the maximum age, $t_{\text {max. }}$. Correlations among biological parameters have been used primarily to obtain empirical formulae to estimate $M$ (Beverton, 1992). Relationships among $\kappa, L_{\infty}$, $T_{m}, L_{m}, M$ and $t_{\max }$ were investigated by Beverton and Holt (1959) who found that they are somewhat invariant across species, and across populations within a species (Jenson, 1996). From such studies, the relationships between; a) $L_{m}, L_{\infty}$ and $M$, b) $L_{\infty}, \kappa$ and $M$, and c) $M$ and $t_{\text {max }}$ became known and empirical formulae were developed to calculate $M$ (e.g. Pauly, 1980; Hoenig, 1983).
More recently, examination of biological parameters and their inter-relationships have been used to assess or categorise the vulnerability of a stock to fishing pressure (e.g. Frisk et al. 2001). In the absence of comprehensive data on the population dynamics of a stock which would allow the application of formal stock assessment methods, 'rapid assessments' are an increasingly important tool that rely on basic lifehistory parameters only. The increasing wealth of information available for well-studied species, however, could give rise to methods of estimating unknown parameters for a 'data-poor' species, from those for 'data-rich' species. While it is more desirable to have direct estimates of life history parameters for each species, the value/importance of some species, especially those that are not target species, may impede funding for the necessary research. Assessments based on data for previously well-studied species potentially offer insights into the biological characteristics of poorly-studied species, necessary for informed management, at relatively small cost.
Information on the biological parameters of fish stocks is often 'locked' away in grey literature, spread out amongst the plethora of literature, or 'hidden' on the hard drives or in the filing cabinets of individual scientists. Myers (2000) addressed this issue by collating information from more than 700 fish stocks made up of over 160 species. The data assembled by Myers (2000) includes not only biological parameters but also spawner-recruit time-series. These data are stored in a format that is easily accessible to external users.

This chapter describes a meta-analysis of production parameters and presents an examination of biological parameters and their inter-relationships.

## Methods

## Species characteristics

The species/stocks considered in this chapter are characterised roughly according to their biological parameters, their diet and depth preferences, and their habitat usage. Characterisations were based on the descriptions of biology from FishBase. Non-numerical quantities (for diet, depth and habitat) are allocated numeric codes to enable rough comparisons to be made among species to allow 'similar' species to be identified automatically, and with relative ease. Reported values were used for all numeric fields.
Values are reported for the following quantities (units, where applicable, in parenthesis):
the maximum age, $t_{\text {max }}$, (years);
the rate of natural mortality, $M\left(\mathrm{yr}^{-1}\right)$;
the parameters of the von Bertalanffy growth equation:

$$
L_{a}=L_{\infty}\left(1-\exp \left(-\kappa\left(a-t_{0}\right)\right)\right)
$$

where the units for $L_{\infty}$ and $t_{0}$ are cm and years respectively;
a) the age at $50 \%$ maturity, $T_{\mathrm{m}}$ (years);
b) the length at $50 \%$ maturity, $L \mathrm{~m}(\mathrm{~cm})$;
c) the extent of variation in recruitment, quantified by the standard deviation of deviations in recruitment about the stock recruitment relationship, $\sigma_{r}$;
d) the steepness of the stock-recruitment relationship (the fraction of the number of recruits to be expected when the spawner biomass is reduced to $20 \%$ of its unfished size), $z$;
e) diet preference ( $1=$ Phytoplankton or zooplankton; $2=$ Salps; $3=$ Invertebrates; $4=$ Invertebrates and fish);
f) depth preference $(1=$ Shelf $(0-200 \mathrm{~m}) ; 2=$ Shelf-slope $(0-700 \mathrm{~m}) ; 3=$ Upper slope $(200-700 \mathrm{~m}) ; 4=$ Lower slope ( $700 \mathrm{~m}+$ ) );
g) habitat usage ( $1=$ Demersal; $2=$ Benthopelagic; $3=$ Pelagic $)$;
h) the latitude of spawning (a positive number means north, a negative number means south);
i) the yield (kg) per recruit at $F=0$;
j) average water temperature;
k) fecundity, and

1) parameters of the length-weight relationship ( $a$ and $b$ ).

These biological parameters have been chosen because of their availability. They were used in further analyses depending on whether:

1) they are well-defined so that most analysts estimate comparably-defined quantities,
2) they are readily available for a large number of fish stocks, and
3) they are important when conducting stock assessments, performing population projections, and for developing fisheries reference points.

## Identification of data sets

Information was extracted from 443 stocks (Table1). Biological characteristics were obtained primarily from Ransom Myers' Stock Recruitment DataBase, the Fishbase website, and the websites of various fisheries institutes. References are included Table 1 as available. Complementary data were obtained from the literature found by searching Current Contents and the Aquatic Sciences and Fisheries Abstracts. Leading authorities on particular species were also contacted by e-mail in an attempt to obtain unpublished or 'grey' literature. Biological parameters of SESSF species are shown in Table 2.

## Biological correlates

The values for the parameters were log-transformed (base 10) and plotted against each other to assess which (if any) of the parameters are able to predict some of the other parameters. The choice of logtransformation was motivated by its use by Pauly (1980) in his model for $M$. The parameters included in this analysis were: a) the parameters of the length-weight relationship ( $a$ and $b$ ), $\mathbf{b}$ ) the asymptotic length $\left.\left(L_{\infty}\right), \mathrm{c}\right)$ the von Bertalanffy growth coefficient $(\kappa), \mathrm{d}$ ) the length-at-maturity ( $L_{\mathrm{m}}$ ) for both sexes and for females only, e) the age-at-maturity ( $T_{\mathrm{m}}$ ) for both sexes and for females only, f ) the maximum age ( $t_{\text {max }}$ ), and f ) the rate of natural mortality $(M)$. All combinations of these parameters were plotted, and axes were reversed in all cases (e.g. $x$ on $y$ as well as $y$ on $x$ ). Duplicates were removed from analyses where the values for all of the parameters examined were equal to reduce the influence of identical data for multiple stocks of the same species. $95 \%$ confidence intervals for the fitted regression line, adjusted $R^{2} \mathrm{~s}$, and probability that the slope of the regression differs from zero were calculated.
The relationships identified by Pauly (1980) and Hoenig (1983) to estimate $M$ were refitted to the data from this study (separately for temperate stocks and for all stocks), and the resulting relationships compared to those of Pauly (1980) and Hoenig (1983). When temperature was not available for a species in a stock area for which a temperature was available for a different species, temperatures were assumed to be the same. When there were two or more different temperatures for given a stock area, the mean of those temperatures was used to specify missing values. When $L_{\infty}$ and $\kappa$ were available for each sex, but not for both sexes combined, the sex-specific values were averaged. In addition, Pauly's (1980) equation was re-fitted to his data where $M$ was equal to or less than 0.5 for comparison with results from the current project.

## Results and Discussion

The two parameters of the length-weight relationship are strongly correlated ( $p<0.001$ ). However, there are several obvious outliers that fall well outside of the $95 \%$ CI for the regression line (Figure 1; Table 3). The relationship between $a$ and $b$ is uninformative because these parameters are estimated together, but this relationship is included to show that the relationship between parameters can become clearer by reversing the axes (Figure 2; Table 4). The parameters $a$ and $b$ are also significantly correlated with $L_{\infty}$, while $b$ is significantly correlated with $L_{m}$. Despite a low $R^{2}$, both $a$ and $b$ are also significantly correlated with $M$.
$L_{\infty}$ is highly correlated with most of the other parameters (Figure 3; Table 5). This result highlights the importance of age data and perhaps that estimating growth parameters should be given priority when researching data-poor fish stocks. The correlation between $L_{m}$ and $L_{\infty}$ is particularly strong, with $L_{\infty}$ explaining $54 \%$ of the variability in $L_{m}$. Frisk et al (2001) also found strong relationships between $L_{m a x}$ and $L_{m}$ ( $L_{\infty}$ was used for $L_{m a x}$ in their study when available).
The other von Bertalanffy parameter examined, $\kappa$, did not correlate significantly with as many parameters as $L_{\infty}$ (Figure 4; Table 6). Not surprisingly given the dependence of $L_{\infty}$ on $\kappa$, $\kappa$ is highly correlated with $T_{m}, t_{\max }$ and $M$. As with $L_{\infty}$, Frisk et al (2001) also found strong relationships between $\kappa$ and $M$ for elasmobranchs. They, however, found that the slope of this relationship (0.42) differed significantly from that of teleosts (1.10). The results in Table 6 show that the slope for teleosts (0.454) in our study is nearly exactly the same as the value found by Frisk et al (2001) for elasmobranches. This was initially thought to be due to the inclusion in the analyses of this chapter of some long-lived species which generally have a lower $M / K$ ratio and hence should reduce the slope of the relationship between $\kappa$ and $M$ (Frisk et al., 2001). However, removal of data for stocks with a $t_{\text {max }}$ greater than 50 years resulted in no change to the slope of the relationship.
Length-at-maturity is highly correlated with age-at-maturity as expected, and also with $t_{\max }$ and several other parameters (Figures 5 and 6; Tables 7 and 8). Age-at-maturity is significantly correlated with $t_{\text {max }}$, and consequently, negatively correlated with $M$ (Figures 7 and 8; Tables 9 and 10). The relationships between $T_{m}, t_{\max }$ and $M$ illustrate the trade-off between growing faster and maturing early and dying at a greater rate, and growing and maturing slower and having lower rate of natural mortality.
$t_{\text {max }}$ is strongly correlated with $M\left(R^{2}=0.60\right)$ with very tight $95 \%$ confidence intervals (Figure 9; Table 11). Plots with $M$ as the independent variable are shown in Figure 10 and Table 12 These results support the use of $t_{\max }$ in the estimation of $M$, showing that $60 \%$ of the variability in $M$ can be explained by $t_{\max }$.
Combinations of parameters were examined to determine if the estimation of $M$ could be improved with the addition of extra parameters into the single parameter models. No improvements in $R^{2}$ were observed (Table 13). However, some large reductions in $R^{2}$ resulted from the reduction in the number of observations compared to the single parameter models. The cause of this was missing one of the parameters for a species when the other parameter was present (i.e. if $L_{\infty}$ but not $T_{m}$ was known for one species, it would have been included in the single parameter analysis based on $L_{\infty}$, but not the multiparameter analysis that used both $L_{\infty}$ and $T_{m}$ ).
The estimates of the coefficients of the relationship between $\log _{10} M$ and $\log _{10} \tau_{\max }$ from this study (all stocks and temperate stocks only) were very similar to those estimated by Hoenig (1983) (Table 14). The $R^{2}$ were high ( 0.64 and 0.61 respectively) for both analyses and the regression lines appear to fit the data reasonably well for the available range of values for $\log _{10} \tau_{\text {max }}$ (Figure 11). In contrast, the coefficients of the relationship between $\log _{10} M$ and $\log _{10} L_{\infty}, \log _{10 \mathrm{~K}}$ and $\log _{10} T$ based on the data in this study differ markedly from those reported by Pauly (1980) (Table 14); specifically, the coefficients were much smaller, particularly when all of the data were used. $R^{2}$ s were also low indicating that the model does not explain much of the variability. The inability to capture much of the variability in $\log _{10} M$ is surprising given that Pauly (1980) reports an $R^{2}$ of 0.85 for his original relationship. Figure 12 plots estimates of $M$ using Pauly's equation and those derived from the data in the current study. The estimates of $M$ based on the regression model of this study are notably smaller than those based on Pauly's model, with the discrepancy increasing with $M$. In the most extreme case, $M$ from our equation was $0.74 \mathrm{yr}^{-1}$ compared to $2.32 \mathrm{yr}^{-1}$ from Pauly's equation; the actual value of $M$ for the stock in question the data set was $1 \mathrm{yr}^{-1}$.

When Pauly's equation was re-fitted to the subset of his original data set for which $M \leq 0.5 \mathrm{yr}^{-1}$, the estimates for $M$ were more similar to those estimated using our equation (Figure 13).
The relationships between $T_{m}$ and $t_{m a x}$, and $T_{m}$ and $M$ for each family from which data were collected are presented in Table 15. Data for other species from previous studies are also included in Table 15 for comparison. The results from this study are comparable with those of the previous studies for all groups, particularly for $T_{m} / t_{\text {max }}$.
Although most of the correlations in Tables 3-13 are weak, there are nevertheless some apparently noteworthy relationships. Specifically, over $50 \%$ of the variability in the length-at-maturity (both sexes) can be explained by $L_{\infty}$ and vice versa (Tables 5 and 7), although the evidence for strong correlation disappears when female length-at-maturity is used instead of the length-at-maturity for both sexes (Tables 5 and 8). There are also strong (and highly statistically significant) correlations between the age-at-maturity and the maximum age (Tables 9 and 11). This correlation is analogous to that between $L_{\infty}$ and $L \mathrm{~m}$. Somewhat surprisingly, there is only a weak correlation between the age-at-maturity and the length-at-maturity (Table 9).
The ability to predict $M$ is possibly of greatest interest to those conducting assessments of 'data-poor' species. The results in Tables 3-13 provide some, but not much, encouragement. Many biological parameters are highly correlated with $M$ (Table 12) although the relationships are generally rather weak. The best explanatory variable for $M$ is the maximum age (Table 12; Figure 11) although considerable uncertainty remains about the relationship between $M$ and $t_{\text {max }}$. A complication particular to predicting $M$ is that independent estimates of this quantity are seldom available so that the values for $M$ in Table 1 are likely guesses or values obtained from the relationships developed by Pauly (1980) or Hoenig (1983). As such, any correlations between $M$ and biological parameters should be interpreted with caution. Nevertheless, the results in Tables 12 and 13 imply that if a way of estimating $M$ based on biological data is needed, the best predictor is $t_{\text {max }}$ (using, for example, Hoenig's (1983) formula or the alternative relationships in Table 14). The weak relationship between $L_{m}$ and $M$ (Table 12) suggests that some ageing data are needed to estimate $M$ (either using $t_{\text {max }}$ or the parameters of the growth curve).
The values for the coefficients estimated for the relationship between $M$ and $t_{\text {max }}$ were similar to those obtained by Hoenig (1983). This is not surprising given that some of the values of $M$ had obviously been obtained using Hoenig's formula. The differences in coefficients observed (smaller constant and larger coefficient for $t_{\max }$ ) are both likely to have resulted from the larger proportion of greater estimates of $t_{\text {max }}$ in our data set (less than $60 \%$ of the Hoenig's samples had a $t_{\text {max }}$ greater than 10 years compared to nearly $70 \%$ in our data set) and because Hoenig used mostly unexploited or lightly exploited stocks in his analyses. Our re-estimation of the coefficients of Pauly's (1980) equation for $M$ revealed very different results, particularly at higher values of $M$. This appears to be caused largely by the high values of $M$ in Pauly's data set. When the equation was re-fitted to his data set for values of $M$ less than or equal to $0.5 \mathrm{yr}^{-1}$, estimates of $M$ were much more similar to those from our equation.
The results of this chapter provide defaults for estimating key biological parameters using estimates of parameters that are easier to determine. However, it is necessary to consider not just the estimates based on the relationships, but also the variability about those relationships. For example, the confidence intervals about the lines in Figures1 - 10 represent the uncertainty about the mean behaviour. The variability for an individual stock will be much larger than implied by these confidence intervals. Therefore, if relationships such as those estimated in this chapter are to be used in stock assessments, sensitivity tests based on the uncertainty of the estimates need to be conducted.

Information for 'data-rich' species to inform assessments of 'data-poor' species

Table 1.Fish stocks examined and their population parameters.



| $\begin{aligned} & 5 \\ & 0 \\ & 0 \\ & 0 \\ & 0 \\ & 0 . \\ & 0 . \\ & 0 \end{aligned}$ | Order | Family | Name | stock |  |  |  | $\frac{\frac{2}{4}}{4}$ | 旁 |  |  | $\begin{aligned} & \frac{\pi}{i} \\ & \frac{1}{2} \\ & \frac{9}{3} \end{aligned}$ |  |  | $\frac{2}{2}$ |  | $\frac{\stackrel{t}{e}}{\stackrel{\rightharpoonup}{e}}$ | $\sim$ | 9 | $\stackrel{\square}{\square}$ | $\stackrel{\square}{\square}$ |  |  | ？ | 3 | $\begin{aligned} & \frac{\pi}{2} \\ & \frac{2}{2} \\ & 0 \end{aligned}$ |  | $\begin{aligned} & \text { 倉 } \\ & \stackrel{\rightharpoonup}{9} \end{aligned}$ | $\begin{aligned} & \frac{\pi}{9} \\ & \frac{8}{4} \\ & \frac{8}{8} \end{aligned}$ |  | $\begin{aligned} & \frac{2}{6} \\ & \frac{8}{8} \end{aligned}$ |  | $\begin{aligned} & 2 \\ & \frac{2}{2} \\ & \frac{8}{8} \end{aligned}$ | 管 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  | Herring（Clupea harengus） | NAFO－4T－（Spring－spawners） | 35.2 |  | 0.880 |  |  |  |  |  |  | ${ }^{4}$ |  |  |  | 0.956 |  | 1 | 1 | 3 | 47.00 | 23 |  |  |  |  | 0.0026 | 3.328 |  |  |  |  |
|  |  |  | Herring（Clupea harengus） | Nafo－4wx | 35.2 |  | 0.280 |  |  |  |  |  |  | 4 |  |  |  | 0.998 |  | ${ }^{1}$ | 1 | 3 | 42.00 | 19 |  |  |  |  | 0.0026 | 3.328 |  |  |  |  |
|  |  |  | Herring（Clupea harengus） | Newfoundland（AB） | 35.2 |  | 0.280 |  |  |  |  |  |  | 5 |  |  |  | 3.564 |  | 1 | 1 | 3 | 50.00 | 19 |  |  |  |  | 0.0026 | 3.328 |  |  |  |  |
|  |  |  | Herring（Clupea harengus） | Archipelago－and－Bothnian－Seas | 21 | －1．16 | 0.380 |  |  |  |  |  |  | 2.5 |  |  |  | 0.465 |  | 1 | 1 | 3 | 61.00 | 17 | 0.2 |  |  |  | 0.0121 | 2778 |  |  |  |  |
|  |  |  | Herring（Clupea harengus） | Bothnian－Bay | 21 | ${ }^{-1.16}$ | 0.380 |  |  |  |  |  |  | 2.5 |  |  |  | 0.685 |  | 1 | 1 | 3 | 65.00 | 17 | 0.15 |  |  |  | 0.0121 | 2778 |  |  |  |  |
|  |  |  | Herring（Clupea harengus） | Central－CosastB．C． | 35.2 |  | 0.880 |  |  |  |  |  |  | 4 |  |  |  |  |  | 1 | 1 | 3 | 52.30 | 19 | 0.373 | 0.25 |  |  | 0.0026 | 3.228 |  |  |  |  |
|  |  |  | Herring（Clupea harengus） | Newfoundand（CD） | ${ }^{35.2}$ |  | 0.280 |  |  |  |  |  |  | 5 |  |  |  | 4.762 |  | 1 | 1 | 3 | 48.45 | 19 |  |  |  |  | 0.0026 | 3.328 |  |  |  |  |
|  |  |  | Herring（Clupea harengus） | Crig－Alaska | 35.2 |  | 0.280 |  |  |  |  |  |  | 3 |  |  |  |  |  | 1 | 1 | 3 | 55.29 | 19 | 0.2 |  |  |  | 0.0026 | 3.328 |  |  |  |  |
| $0$ |  |  | Herring（Clupea harengus） | Downsstock | 29.2 |  | ${ }^{0.48}$ |  |  |  |  |  |  | 3 |  |  |  | 2.033 |  | ${ }^{1}$ | 1 | 3 | 52.00 | ${ }^{13}$ | 0.2 | 0.79 | 9 |  | 0.005 | ${ }^{3} 1$ |  |  |  |  |
| $\stackrel{\rightharpoonup}{\underset{\sim}{7}}$ |  |  | Herring（Clupea harengus） | Gulfof－Evina，－White－Sea | 21 | －1．16 | 0.380 |  |  |  |  |  |  | 2.5 |  |  |  | ${ }^{0.464}$ |  | 1 | 1 | 3 | 65.00 | 17 |  |  |  |  | 0.01212 | 2778 |  |  |  |  |
| $\stackrel{\rightharpoonup}{2}$ |  |  | Herring（Clupea harengus） | Easter－Bering－Sea | 21 | －1．16 | 0.380 |  |  |  |  |  |  | 4 |  |  |  |  |  | 1 | 1 | 3 | 58.00 | 17 |  | 0.512 |  |  | 0.0121 | 2778 |  |  |  |  |
| $\frac{0}{0}$ |  |  | Herring（Clupea harengus） | Newfoundland（EF） | 35.2 |  | 0.280 |  |  |  |  |  |  | 5 |  |  |  | 4.272 |  | 1 | 1 | 3 | 47.30 | 19 |  |  |  |  | 0.0026 | 3.328 |  |  |  |  |
| 兩 |  |  | Herring（Clupea harengus） | Georges－Bank | 35.2 |  | 0.880 |  |  |  |  |  |  | ${ }^{3.5}$ |  |  |  | 1.948 |  | 1 | 1 | 3 | 41.30 | 19 | 0.2 | 0.91 |  |  | 0.00263 | 3.328 |  |  |  |  |
| O. |  |  | Herring（Clupea harengus） | Gull－of－Finland | 21 | $-1.16$ | 0.380 |  |  |  |  |  |  | 2 |  |  |  | 0.683 |  | ${ }^{1}$ | 1 | ${ }^{3}$ | 61.00 | 17 | 0.2 | 0.2 |  |  | ${ }^{0.0121}$ | 2778 |  |  |  |  |
| 雨 |  |  | Herring（Clupea harengus） | Newfoundland（GH） | 35.2 |  | 0.880 |  |  |  |  |  |  | 5 |  |  |  | 2.847 |  | ${ }^{1}$ | ${ }^{1}$ | 3 | 46.40 | 19 |  |  |  |  | 0.0026 | 3.228 |  |  |  |  |
| $\begin{aligned} & 5 \\ & 0 \\ & 0 \end{aligned}$ |  |  | Herring（Clupea harengus） | Gulfoomaine | 35.2 |  | 0.280 |  |  |  |  |  |  | 3.5 |  |  |  | 0.993 |  | 1 | 1 | 3 | 43.00 | 22 | 0.2 | 0.68 |  |  | 0.0026 | 3328 |  |  |  |  |
| $10$ |  |  | Herring（Clupea harengus） | Gulfor．eriga | 21 | $-1.16$ | 0．3s0 |  |  |  |  |  |  | 2 |  |  |  | 0.556 |  | 1 | 1 | 3 | 57．30 | 17 | 0.2 | 0.14 |  |  | 0.0121 | 2.778 |  |  |  |  |
| $\begin{aligned} & 3 \\ & 0 \\ & 0 \end{aligned}$ |  |  | Herring（Clupea harengus） | Hokkaid |  |  |  |  |  |  |  |  |  |  |  |  |  | 5.713 |  | 1 | 1 | 3 | 44.30 |  |  |  |  |  |  |  |  |  |  |  |
| 产 |  |  | Herring（Clupea harengus） | Newfoundand（I） | 35.2 |  | 0.280 |  |  |  |  |  |  | 5 |  |  |  | ${ }^{3.45}$ |  | 1 | 1 | 3 | 47.25 | 19 |  |  |  |  | 0.0026 | 3.328 |  |  |  |  |
| $0$ |  |  | Herring（Clupea harengus） | Healand－（Spring－spawners） | 21 | $-1.16$ | 0.380 |  |  |  |  |  |  | 4 |  |  |  | ${ }^{13.55}$ |  | 1 | 1 | 3 | 63.00 | 19 | ${ }_{0} 0.1$ | 2.88 |  |  | 0.0121 | 2778 |  |  |  |  |
| － |  |  | Herring（Clupea harengus） | Iceland－（Summer－spawners） | 21 | $-1.16$ | 0.380 |  |  |  |  |  |  | ${ }^{3} 5$ |  |  |  | ${ }^{1.141}$ |  | 1 | ${ }^{1}$ | ${ }^{3}$ | 63.00 | 18 | ${ }^{0.1}$ | 2.13 |  |  | 0.0121 | 2.778 |  |  |  |  |
| $\bigcirc$ |  |  | Herring（Cupea harengus） | Gulfof：－Kandalaksha，－White－sea | ${ }^{21}$ | $-1.16$ | 0.380 |  |  |  |  |  |  | 2.5 |  |  |  | 1.601 |  | ${ }^{1}$ | 1 | 3 | 66.24 | 17 |  |  |  |  | 0.01212 | 2778 |  |  |  |  |
| 0 |  |  | Herring（Clupea harengus） | Northem－Trish－Sea | 29.5 |  | ${ }^{0.390}$ |  |  |  |  |  |  | 3 |  |  |  | ${ }^{0.4}$ |  | 1 | ${ }^{1}$ | ${ }^{3}$ | 54.05 | 12 | 0.2 | 0.54 | ${ }^{12}$ |  | 0.0050 | 3.92 |  |  |  |  |
| $\frac{\partial}{\partial}$ |  |  | Herring（Clupea harengus） | Norway－（Spring－spawners） | ${ }^{21}$ | ${ }^{-1.16}$ | 0.380 |  |  |  |  |  |  | 4.5 |  |  |  | ${ }^{13.77}$ |  | 1 | 1 | ${ }^{3}$ | 65.00 | 23 | 0.13 | 1.62 |  |  | 0.0075 | ${ }^{3.0}$ |  |  |  |  |








$\frac{\text { Information for 'data-rich' species to inform assessments of 'data-poor' species }}{20}$


$\frac{\text { Information for 'data-rich' species to inform assessments of 'data-poor' species }}{22}$

\＆乙



| Order | Family | Name | stock | $\begin{aligned} & \frac{\pi}{巳} \\ & \frac{5}{5} \\ & \frac{5}{6} \\ & \hline \end{aligned}$ |  |  | $\frac{3}{\frac{3}{2}}$ | $\stackrel{3}{\square}$ | 輁 | 范 |  |  |  | $\begin{aligned} & \frac{\pi}{2} \\ & \frac{2}{2} \\ & \frac{2}{2} \end{aligned}$ |  |  | $\sim$ | 9 | $\stackrel{\square}{\square}$ | $\stackrel{\square}{\square}$ |  |  | ？ | 3 | $\begin{aligned} & \frac{\pi}{2} \\ & \frac{\hat{2}}{2} \\ & \frac{D_{0}}{2} \end{aligned}$ |  |  | $\begin{aligned} & \frac{\pi}{0} \\ & \frac{2}{4} \\ & \frac{8}{8} \end{aligned}$ |  | $\begin{aligned} & \frac{2}{6} \\ & \frac{8}{8} \end{aligned}$ | 年 | 翣 | 管 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Mackere（Scomber scombrus） | Black－sea | ${ }^{41.0}$ |  | ${ }^{0.4}$ |  |  |  |  |  |  | 2 |  | ${ }^{26}$ |  | 1.850 |  | ${ }^{4}$ | 1 | 3 | 40.30 | 17 | 1.17 |  |  |  | 0.046 | 3.18 |  |  |  |  |
|  |  | Mackerel（Scomber scombrus） | Wester－ICES | 33.6 | －0．18 | 0.837 |  |  |  |  |  |  | ${ }^{3}$ |  | 32 |  | 0.71 |  | 4 | 1 | ${ }^{3}$ | 49.00 | 17 | 0.15 | 1.87 | 9 |  | 0.0046 | 3.18 |  |  |  |  |
|  |  | Kingmackerel（Somberomorus cavala） | W．Gulfor－Mexico | 115 | 0.256 | 0.1658 |  |  |  |  |  |  |  |  |  | 47 | 2889 |  | ${ }^{4}$ | 1 | 3 | 21.00 | 14 | 0.1658 |  |  |  | 0.0150 | 2.893 |  |  |  |  |
|  |  | Albacore－tuna（Thunnus alalunga） | South－Pacific－Cean | 127.0 | －231 | 0.12 |  |  |  |  |  |  |  | 5 | 80 |  | 0.467 |  | 4 | 3 | 3 | －17．00 | 10 | 0.22 |  |  |  | 0.0453 | 279 |  |  |  |  |
|  |  | Yellowfin－tuna（Thumnus albacares） | Eastern－Pacific－Cean | 19.10 | －1．02 | 0.327 |  |  |  |  |  |  | 2 |  |  |  | 0.293 |  | ${ }^{4}$ | 1 | 3 | 10.00 | 8 | 0.8 | ${ }_{8.5}$ | 22 |  | 0.0216 | 2.981 |  |  |  |  |
|  |  | Yelowfin－tuna（Thumus albacares） | Indian－Ocean | 19.0 |  | ${ }^{0.163}$ |  |  |  |  |  |  | 2 |  |  |  | 0.846 |  | ${ }_{4}$ | 1 | 3 | －10．00 | 8 | 0.4 |  | 25 |  | 0.0514 | 2.858 |  |  |  |  |
|  |  | Southem－bluefin－tuna（Thurnus maccoyi） | Southem－Pacific | 220 |  | 0.140 |  |  |  |  |  |  | 8 |  | 125 |  | 0.238 |  | ${ }_{4}$ | 1 | 3 | －15．00 | 20 | 0.08 | 180 |  |  | 0.0265 | 2.94 |  |  |  |  |
|  |  | Southem－bluefin－tua（Thunnus maccoyi） | Southem－Pacific | 220 |  | 0.140 |  |  |  |  |  |  | 8 |  | 125 |  | 0.473 |  |  |  |  | －15．00 | 20 | 0.08 | ${ }_{180}$ |  |  | 0.0265 | 2.94 |  |  |  |  |
|  |  | Bigeye－Tuna（Thunnus obesus） | West－Atantic | 219.0 |  | 0.23 |  |  |  |  |  |  | ${ }^{4}$ |  | 100 |  | 0.201 |  | ${ }^{4}$ | ${ }^{4}$ | 3 | －20．00 | 11 | 0.4 | 15.5 | 20 |  | 0.0125 | 3.121 |  |  |  |  |
|  |  | Bigeye－Tuna（Thurnus obesus） | Eastracific | 187.0 |  | 0.380 |  |  |  |  |  |  | 3 |  |  |  | 0.104 |  | ${ }_{4}$ | ${ }_{4}$ | 3 | －10．00 | 11 | 0.6 | 22 | 23 |  | 0.0178 | 2.902 |  |  |  |  |
|  |  | Bigeye－Tuna（Thunnus obesus） | East－Pacific | 187.0 |  | 0.380 |  |  |  |  |  |  | ${ }^{3}$ |  |  |  | 0.104 |  | 4 | 4 | 3 | －10．00 | 11 | 0.44 | 425373 | 23 |  | 0.0178 | 2.902 |  |  |  |  |
|  |  | Atantic－bluefin－tuna（Thurnus thynnus） | WestAAlantic |  |  |  | ${ }^{278}$ |  | 0.170 | 266.0 |  | 0.170 | s |  |  |  | 0.966 |  | ${ }^{4}$ | 1 | 3 | 20.00 | 15 | 0.14 | 940.72 | 15 |  | 0.0904 | 2.837 |  |  |  |  |
|  |  | Sea－bream（Chrysophrys major） | Yelow－Sea | 70.8 |  | 0.117 |  |  |  |  |  |  | 3 |  |  |  | 1.392 |  | ${ }^{4}$ | 1 | 1 | 35.00 | 20 | 0.357 |  |  |  | 0.0199 | 3 |  |  |  |  |
|  |  | Snapper（Pagrus auratus） | New－Zealand，SNA－8 | 66.7 | －0．11 | 0.16 |  |  |  |  |  |  | ${ }^{4}$ |  | 25 |  | 0.805 |  | ${ }^{4}$ | 2 | 2 | －37．00 | 35 | 0.06 | 50．956 |  |  | 0.0446 | 2793 |  |  |  |  |
|  |  | Snapper（Pagrus auratus） | Haurak－Gulffay－of：Penty | 58.8 | －1．11 | 0.102 |  |  |  |  |  |  | ${ }^{4.5}$ |  | 25 |  | 0.749 |  | ${ }_{4}$ | 2 | 2 | －3600 | 35 | 0.06 | 26.574 |  |  | 0.0467 | 2793 |  |  |  |  |
|  |  | Snapper（Pagrus auratus） | NZ，SNA |  |  |  | 58.8 |  |  |  |  | 0.102 |  | 4 |  | ${ }^{24}$ |  | 0.94 | 3 | 1 | 1 |  | ${ }_{60}$ | 0.075 |  |  |  |  |  |  |  |  |  |
|  |  | Snapper（Pagrus auratus） | NZ．SNA－S |  |  |  | ${ }_{66} 7$ | －0．11 | 0.16 | 66.7 | －0．11 | 0.16 |  | 4 |  | 24 | 0.62 | 0.94 | 3 | 1 | 1 |  | ${ }^{60}$ | 0.075 |  |  |  |  |  |  |  |  |  |
|  |  | Red－porgy（Pagrus pagrus） | North－Carolina |  |  |  | 76.3 | －9999 | 0.096 | 76.3 | －9999 | 0.096 |  | 4 |  | 23 | 0.4 |  | 4 | 1 | 2 | 35.00 | 17 |  | $0.59794$ |  |  | 0.0102 | 3.06 |  |  |  |  |
|  |  | Scup（Stenotomus cryssops） | CapeCod－CApe－HaterassUSA | ${ }_{42} 4$ |  | 0.17 |  |  |  |  |  |  | 2 |  |  | ${ }^{16}$ | 0.32 |  | 3 | 1 | 1 | 39.00 | 15 | 0.2 | 2.5439 |  |  |  |  |  |  |  |  |
|  |  | Yelow－saa－bream（Taius tumifrons） | Central－East－China－Sea | 47.4 | －0．3 | 0.249 |  |  |  |  |  |  | 3 |  |  |  | 0.532 |  | ${ }^{4}$ | 1 | 2 | 30.00 |  |  |  |  |  |  |  |  |  |  |  |
|  |  | Yelow－sa－bream（Taius tumifirons） | East－China－Sea | 47.4 | －0．3 | 0.249 |  |  |  |  |  |  | 3 |  |  |  | 0.222 |  | 4 | 1 | 2 | 30.00 |  | 0.2 |  |  |  |  |  |  |  |  |  |





$\frac{\text { Information for 'data-rich' species to inform assessments of 'data-poor' species }}{30}$


Information for 'data-rich' species to inform assessments of 'data-poor' species
Table 2. Population parameters for SESSF species.



Table 3. Regression statistics of relationships between biological parameters with as the independent variable.

| Model | $R^{2}{ }_{\text {adj }}$ | $p_{\text {overall }}$ | RMSE | $n$ |
| :---: | :---: | :---: | :---: | :---: |
| $\log _{10} b=-0.036(0.004) \times \log _{10} a+0.412$ (0.009) | $\underline{0.3263}$ | $\bigcirc 0.0001$ | 0.0231 | 169 |
| $\log _{10} L_{\infty}=0.112(0.058) \times \log _{10} a+1.978$ (0.126) | $\underline{0.0190}$ | $\underline{0.0550}$ | 0.3007 | 142 |
| $\log _{10} \kappa=-0.062(0.051) \times \log _{10} a-0.756$ (0.112) | 0.0032 | 0.2294 | 0.2657 | 142 |
| $\log _{10} L_{\mathrm{m}}=0.098(0.124) \times \log _{10} a+1.616$ (0.272) | -0.0119 | 0.4355 | 0.3466 | 32 |
| $\log _{10} L_{\mathrm{m}} \mathrm{f}=-0.022(0.077) \times \log _{10} a+1.492$ (0.172) | -0.0367 | 0.7805 | 0.1789 | 26 |
| $\log _{10} T_{\mathrm{m}}=0.029$ (0.039) $\times \log _{10} a+0.589$ (0.085) | -0.0032 | 0.4585 | 0.2042 | 139 |
| $\log _{10} T_{\mathrm{m}} \mathrm{f}=-0.094(0.093) \times \log _{10} a+0.378$ (0.212) | 0.0009 | 0.3219 | 0.2660 | 25 |
| $\log _{10} t_{\text {max }}=-0.007(0.043) \times \log _{10} a+1.226$ (0.094) | -0.0061 | 0.8743 | 0.2466 | 160 |
| $\log _{10} M=-0.112$ (0.053) $\times \log _{10} a-0.845$ (0.115) | $\underline{0.0231}$ | $\underline{0.0361}$ | 0.2698 | 147 |

All analyses used $\log _{10}$ transformed data. Regression coefficients are shown as $\pm 1 \mathrm{SE}$ in parenthesis; $R^{2}{ }^{2}$ adj $=$ correlation coeficient; $p=$ probablility level; RMSE = root mean square error from regression model, and $n=$ number of stocks (_f after parameter indicates that parameter was estimated for females only). Results for which $p<0.1$ are indicated by underlines and those for which $p<0.05$ by double-underlines.

Table 4. Regression statistics of relationships between biological parameters with $b$ as the independent variable.

| Model | $R^{2}$ adj | $p_{\text {overall }}$ | RMSE | $n$ |
| :---: | :---: | :---: | :---: | :---: |
| $\log _{10} a=-9.173$ (1.008) $\times \log _{10} b+2.349$ (0.493) | $\underline{0.3263}$ | $\bigcirc 0.0001$ | 0.3680 | 169 |
| $\log _{10} L_{\infty}=-2.505(0.854) \times \log _{10} b+2.960$ (0.418) | $\underline{\underline{0.0504}}$ | $\underline{\underline{0.0039}}$ | 0.2960 | 143 |
| $\log _{10} \kappa=1.071$ (0.764) $\times \log _{10} b-1.147$ (0.374) | 0.0067 | 0.1630 | 0.2647 | 143 |
| $\log _{10} L_{\text {m }}=-4.318$ (1.949) $\times \log _{10} b+3.516$ (0.954) | $\underline{\underline{0.1088}}$ | $\underline{\underline{0.0342}}$ | 0.3253 | 32 |
| $\log _{10} L_{\mathrm{m}} \mathrm{f}=-0.832(1.380) \times \log _{10} b+1.937$ (0.680) | -0.0241 | 0.5515 | 0.1879 | 27 |
| $\log _{10} T_{\mathrm{m}}=-0.809(0.630) \times \log _{10} b+0.921$ (0.308) | 0.0046 | 0.2012 | 0.2036 | 140 |
| $\log _{10} T_{\mathrm{m}} \mathrm{f}=2.780(1.967) \times \log _{10} b-0.805$ (0.970) | 0.0370 | 0.1699 | 0.2790 | 26 |
| $\log _{10} t_{\text {max }}=-1.190$ (0.698) $\times \log _{10} b+1.819$ (0.342) | $\underline{\underline{0.0117}}$ | $\underline{\underline{0.0904}}$ | 0.2463 | 161 |
| $\log _{10} M=3.162$ (0.792) x Log $10 ~ b-2.146 ~(0.387) ~$ | $\underline{\underline{0.0916}}$ | $\underline{\underline{0.0001}}$ | 0.2641 | 148 |

All analyses used $\log _{10}$ transformed data. Regression coefficients are shown as $\pm 1 \mathrm{SE}$ in parenthesis; $R^{2}{ }_{\text {adj }}=$ correlation coeficient; $p=$ probablility level; RMSE = root mean square error from regression model, and $n=$ number of stocks (_f after parameter indicates that parameter was estimated for females only). Results for which $p<0.1$ are indicated by underlines and those for which $p<0.05$ by double-underlines.

Table 5. Regression statistics of relationships between biological parameters with $L_{\infty}$ as the independent variable.

| Model | $R^{2}$ adj | $p_{\text {overall }}$ | RMSE | $n$ |
| :---: | :---: | :---: | :---: | :---: |
| $\log _{10} a=0.231$ (0.120) $\times \log _{10} L_{\infty}-2.530$ (0.210) | $\underline{0.0190}$ | $\underline{\underline{0.0550}}$ | 0.4314 | 142 |
| $\log _{10} b=-0.023$ (0.008) $\times$ Log ${ }_{10} L_{\infty}+0.528$ (0.014) | $\underline{\underline{0.0504}}$ | $\underline{0.0039}$ | 0.0282 | 143 |
| $\log _{10} \kappa=-0.618(0.061) \times \log _{10} L_{\infty}+0.444$ (0.106) | $\underline{0.3634}$ | $\bigcirc .0001$ | 0.2380 | 181 |
| $\log _{10} L_{\mathrm{m}}=0.836$ (0.135) $\times \log _{10} L_{\infty}+0.001$ (0.223) | $\underline{0.5382}$ | $\bigcirc 00001$ | 0.2201 | 32 |
| $\log _{10} L_{\text {m_ }} \mathrm{f}=0.439(0.149) \times \log _{10} L_{\infty}+0.766$ (0.267) | $\underline{\underline{0.2097}}$ | $\underline{0.0064}$ | 0.1778 | 29 |
| $\log _{10} T_{\mathrm{m}}=0.343(0.057) \times \log _{10} L_{\infty}-0.070$ (0.099) | $\underline{0.1874}$ | ¢. 0001 | 0.2101 | 155 |
| $\log _{10} T_{\mathrm{m} /} \mathrm{f}=0.670$ (0.270) $\times$ Log ${ }_{10} L_{\alpha_{\infty}-0.550}(0.501)$ | $\underline{0.1554}$ | $\underline{0.0197}$ | 0.2963 | 28 |
| $\log _{10} t_{\text {max }}=0.343(0.063) \times \log _{10} L_{\infty}++0.636$ (0.111) | $\underline{\underline{0.1507}}$ | $\leq .0001$ | 0.2412 | 161 |
| $\log _{10} M=-0.297$ (0.079) $\times \log _{10} L_{\infty}-0.064$ (0.139) | 0.0764 | $\underline{0.0002}$ | 0.2996 | 158 |

All analyses used $\log _{10}$ transformed data. Regression coefficients are shown as $\pm 1$ SE in parenthesis; $R^{2}{ }_{\mathrm{adj}}=$ correlation coeficient; $p=$ probablility level; RMSE = root mean square error from regression model, and $n=$ number of stocks (_f after parameter indicates that parameter was estimated for females only). Results for which $p<0.1$ are indicated by underlines and those for which $p<0.05$ by double-underlines.

Table 6. Regression statistics of relationships between biological parameters with $\kappa$ as the independent variable.

| Model | $R^{2}$ adj | $p_{\text {overall }}$ | RMSE | $n$ |
| :---: | :---: | :---: | :---: | :---: |
| $\log _{10} a=-0.166$ (0.137) $\times \log _{10} \kappa-2.24$ (0.093) | 0.0032 | 0.2294 | 0.4348 | 142 |
| $\log _{10} b=0.013$ (0.009) $\times \log _{10} \kappa+0.497$ (0.006) | 0.0067 | 0.1630 | 0.0289 | 143 |
| $\log _{10} L_{\infty}=-0.593(0.058) \times \log _{10} \kappa+1.355$ (0.040) | $\underline{\underline{0.3634}}$ | $\bigcirc .0001$ | 0.2332 | 181 |
| $\log _{10} L_{m}=-0.428(0.179) \times \log _{10} \kappa+1.139(0.107)$ | $\underline{\underline{0.1282}}$ | $\underline{\underline{0.0232}}$ | 0.3024 | 32 |
| $\log _{10} L_{\mathrm{m}} \mathrm{f}=0.012(0.101) \times \log _{10} \kappa+1.558$ (0.084) | -0.0340 | 0.9042 | 0.2000 | 30 |
| $\log _{10} T_{\mathrm{m}}=-0.429$ (0.055) $\times \log _{10} \kappa+0.257$ (0.037) | $\underline{\underline{0.2752}}$ | $\bigcirc .0001$ | 0.1982 | 156 |
| $\log _{10} T_{\mathrm{m}} \mathrm{f}=-0.449(0.162) \times \log _{10} \kappa+0.327$ (0.139) | $\underline{\underline{0.1881}}$ | $\underline{\underline{0.0096}}$ | 0.2855 | 29 |
| $\log _{10} t_{\text {max }}=-0.482(0.061) \times \log _{10} \kappa+0.924$ (0.043) | $\underline{\underline{0.2751}}$ | $\bigcirc$ | 0.2250 | 162 |
| $\log _{10} M=0.454(0.073) \times \log _{10} \kappa-0.300$ (0.050) | $\underline{0.1913}$ | $\leq .0001$ | 0.2800 | 159 |

All analyses used $\log _{10}$ transformed data. Regression coefficients are shown as $\pm 1$ SE in parenthesis; $R^{2}{ }^{\text {adj }}=$ correlation coeficient; $p=$ probablility level; RMSE = root mean square error from regression model, and $n=$ number of stocks (_f after parameter indicates that parameter was estimated for females only). Results for which $p<0.1$ are indicated by underlines and those for which $p<0.05$ by double-underlines.

Table 7. Regression statistics of relationships between biological parameters with Lm (sexes combined) as the independent variable.

| Model | $R^{2}$ adj | $p_{\text {overall }}$ | RMSE | $n$ |
| :---: | :---: | :---: | :---: | :---: |
| $\log _{10} a=0.201$ (0.254) x Log ${ }_{10} L_{\text {m-2.414 (0.368) }}$ | -0.0119 | 0.4355 | 0.4956 | 32 |
| $\log _{10} b=-0.032(0.014) \times \log _{10} L_{\text {m }}+0.533$ (0.021) | $\underline{0.1088}$ | $\underline{0.0342}$ | 0.0279 | 32 |
| $\log _{10} L_{\infty}=0.661(0.107) \times \log _{10} L_{m}+0.727(0.149)$ | $\underline{0.5382}$ | <. 0001 | 0.1958 | 32 |
| $\log _{10} \kappa=-0.363$ (0.152) $\times$ Log ${ }_{10} L_{m-0} 0.023$ (0.212) | $\underline{0.1282}$ | $\underline{0.0232}$ | 0.2783 | 32 |
| $\log _{10} T_{\mathrm{m}}=0.386$ (0.119) $\times$ Log ${ }_{10} L_{m-0.124 ~(0.171) ~}^{\text {a }}$ | $\underline{0.2461}$ | $\underline{0.0031}$ | 0.2189 | 29 |
| $\log _{10} t_{\text {max }}=0.359(0.162) \times \log _{10} L_{\text {m }}+0.587(0.238)$ | $\underline{0.1050}$ | $\underline{\underline{0.0346}}$ | 0.3118 | 33 |
| $\log _{10} M=-0.334(0.176) \times \log _{10} L_{m-0} 0.022$ (0.249) | $\underline{0.0711}$ | 0.0665 | 0.3921 | 34 |

All analyses used $\log _{10}$ transformed data. Regression coefficients are shown as $\pm 1$ SE in parenthesis; $R^{2}{ }_{\text {adj }}=$ correlation coeficient; $p=$ probablility level; RMSE = root mean square error from regression model, and $n=$ number of stocks. Results for which $p<0.1$ are indicated by underlines and those for which $p<0.05$ by double-underlines.

Table 8. Regression statistics of relationships between biological parameters with $L_{m}$ (females only) as the independent variable.

| Model | $R^{2}$ adj | $p_{\text {overall }}$ | RMSE | $n$ |
| :---: | :---: | :---: | :---: | :---: |
| $\log _{10} a=-0.146$ (0.518) $\times$ Log ${ }_{10} L_{\mathrm{m}} \mathrm{f}-1.963$ (0.802) | -0.0367 | 0.7805 | 0.4639 | 26 |
| $\log _{10} b=-0.017$ (0.028) x $\log _{10} L_{\text {m_f }} \mathrm{f}+0.518$ (0.042) | -0.0241 | 0.5515 | 0.0265 | 27 |
| $\log _{10} L_{\infty}=0.540$ (0.183) $\times \log _{10} L_{\text {m }} \mathrm{f}+0.945$ (0.286) | $\underline{0.2097}$ | $\bigcirc 0.0001$ | 0.1973 | 29 |
| $\log _{10} \kappa=0.041$ (0.341) x $\log _{10} L_{m \_} \mathrm{f}-0.820$ (0.532) | -0.0340 | 0.9042 | 0.3671 | 30 |
| $\log _{10} T_{\mathrm{m} \_} \mathrm{f}=0.560$ (0.169) $\times \log _{10} L_{\mathrm{m}} \mathrm{f}-\mathrm{f}-216$ (0.264) | $\underline{0.1297}$ | $\leq 0.0001$ | 0.2615 | 67 |
| $\log _{10} t_{\text {max }}=0.365$ (0.182) $\times \log _{10} L_{\mathrm{m}} \mathrm{f}+0.774$ (0.284) | $\underline{0.0383}$ | $\bigcirc 0.0001$ | 0.3051 | 76 |
| $\log _{10} M=-0.321$ (0.186) x Log 10 Lm_f ${ }^{\text {d }} 0.161$ (0.293) | $\underline{0.0234}$ | 0.0891 | 0.3068 | 82 |

All analyses used $\log _{10}$ transformed data. Regression coefficients are shown as $\pm 1$ SE in parenthesis; $R^{2}{ }_{\text {adj }}=$ correlation coeficient; $p=$ probablility level; RMSE = root mean square error from regression model, and $n=$ number of stocks (_f after parameter indicates that parameter was estimated for females only). Results for which $p<0.1$ are indicated by underlines and those for which $p<0.05$ by double-underlines.

Table 9. Regression statistics of relationships between biological parameters with $T_{m}$ (sexes combined) as the independent variable.

| Model | $R^{2}$ adj | $p_{\text {overall }}$ | RMSE | $n$ |
| :---: | :---: | :---: | :---: | :---: |
| $\log _{10} a=0.137(0.185) \times \log _{10} T_{\mathrm{m}}-2.206$ (0.104) | -0.0032 | 0.4585 | 0.4444 | 139 |
| $\log _{10} b=-0.015$ (0.011) $\times \log _{10} T_{\mathrm{m}}+0.497$ (0.006) | 0.0046 | 0.2012 | 0.0273 | 140 |
| $\log _{10} L_{\infty}=0.562(0.093) \times \log _{10} T_{\mathrm{m}}+1.426$ (0.053) | $\underline{\underline{0.1874}}$ | $\bigcirc 0.0001$ | 0.2689 | 155 |
| $\log _{10} \kappa=-0.652(0.084) \times \log _{10} T_{\mathrm{m}}-0.275$ (0.048) | $\underline{\underline{0.2752}}$ | $\bigcirc 0.0001$ | 0.2444 | 156 |
| $\log _{10} L_{\mathrm{m}}=0.705$ (0.218) $\times \log _{10} T_{\mathrm{m}}+1.100$ (0.105) | $\underline{0.2461}$ | $\leq 0.0001$ | 0.2959 | 29 |
| $\log _{10} t_{\text {max }}=0.740$ (0.070) $\times \log _{10} T_{\mathrm{m}}+0.846$ (0.042) | $\underline{\underline{0.3891}}$ | $\bigcirc 0.0001$ | 0.2072 | 172 |
| $\log _{10} M=-0.772$ (0.075) $\times \log _{10} T_{\mathrm{m}}-0.188$ (0.044) | 0.3685 | $\leq 0.0001$ | 0.2517 | 181 |

All analyses used $\log _{10}$ transformed data. Regression coefficients are shown as $\pm 1 \mathrm{SE}$ in parenthesis; $R^{2}{ }_{\text {adj }}=$ correlation coeficient; $p=$ probablility level; RMSE = root mean square error from regression model, and $n=$ number of stocks. Results for which $p<0.1$ are indicated by underlines and those for which $p<0.05$ by double-underlines.

Table 10.Regression statistics of relationships between biological parameters with $T_{m}$ (females only) as the independent variable.

| Model | $R^{2}$ adj | $p_{\text {overall }}$ | RMSE | $n$ |
| :---: | :---: | :---: | :---: | :---: |
| $\log _{10} a=-0.436$ (0.431) $\times \log _{10} \mathrm{~T}_{\mathrm{m}} \mathrm{f}-1.964$ (0.277) | 0.0009 | 0.3219 | 0.5738 | 25 |
| $\log _{10} b=0.027$ (0.019) $\times \log _{10} \mathrm{~T}_{\mathrm{m}} \mathrm{f}+0.477$ (0.012) | 0.0370 | 0.1699 | 0.0273 | 26 |
| $\log _{10} L_{\infty}=0.277$ (0.112) $\times \log _{10} \mathrm{~T}_{\mathrm{m}} \mathrm{f}+1.652(0.084)$ | $\underline{\underline{0.1554}}$ | $\bigcirc 0.0001$ | 0.1904 | 28 |
| $\log _{10} \kappa=-0.481$ (0.173) $\times$ Log ${ }_{10} \mathrm{~T}_{\mathrm{m}-} \mathrm{f}-0.470$ (0.130) | $\underline{0.1881}$ | $\bigcirc 0.0001$ | 0.2956 | 29 |
| $\log _{10} L_{\mathrm{m}}^{2} \mathrm{f}=0.255$ (0.077) $\times \log _{10} \mathrm{~T}_{\mathrm{m} \_} \mathrm{f}+1.386$ (0.055) | $\underline{0.1297}$ | $\bigcirc 0.0001$ | 0.1764 | 67 |
| $\log _{10} t_{\text {max }}=0.735$ (0.090 ) x $\log _{10} \mathrm{~T}_{\mathrm{m}} \mathrm{f} \mathrm{f}+0.882$ (0.067) | $\underline{\underline{0.4785}}$ | $\bigcirc 0.0001$ | 0.2331 | 71 |
| $\log _{10} M=-0.587$ (0.094) $\times \log _{10} T_{\mathrm{m}} \mathrm{f}$ f-0.304 (0.069) | $\underline{\underline{0.3285}}$ | $\leq 0.0001$ | 0.2464 | 77 |

All analyses used $\log _{10}$ transformed data. Regression coefficients are shown as $\pm 1 \mathrm{SE}$ in parenthesis; $R^{2}{ }_{\text {adj }}=$ correlation coeficient; $p=$ probablility level; RMSE = root mean square error from regression model, and $n=$ number of stocks ( $\_$f after parameter indicates that parameter was estimated for females only). Results for which $p<0.1$ are indicated by underlines and those for which $p<0.05$ by double-underlines.

Table 11.Regression statistics of relationships between biological parameters with $t_{\max }$ as the independent variable.

| Model | $R^{2}{ }_{\text {adj }}$ | poverall | RMSE | n |
| :---: | :---: | :---: | :---: | :---: |
| $\log _{10} a=-0.023$ (0.145) $\times \log _{10} t_{\text {max }}-2.096$ (0.184) | -0.0061 | 0.8743 | 0.4517 | 160 |
| $\log _{10} b=-0.015$ (0.009) $\times \log _{10} t_{\text {max }}+0.507$ (0.011) | $\underline{0.0117}$ | $\underline{0.0904}$ | 0.0276 | 161 |
| $\log _{10} L_{\infty}=0.455(0.084) \times \log _{10} t_{\text {max }}+1.173$ (0.105) | $\underline{\underline{0.1507}}$ | $\bigcirc 0.0001$ | 0.2780 | 161 |
| $\log _{10} \kappa=-0.579$ (0.073) $\times \log _{10} t_{\text {max }}+0.074$ (0.092) | 0.2751 | $\bigcirc 0.0001$ | 0.2466 | 162 |
| $\log _{10} L_{\mathrm{m}}=0.368$ (0.167) $\times \log _{10} t_{\text {max }}+1.020$ (0.191) | $\underline{0.1050}$ | $\bigcirc 0.0001$ | 0.3160 | 33 |
| $\log _{10} L_{\mathrm{m}} \mathrm{f}=0.140$ (0.070) $\times \log _{10} t_{\text {max }}+1.363$ (0.096) | $\underline{0.0383}$ | $\bigcirc 0.0001$ | 0.1886 | 76 |
| $\log _{10} T_{\mathrm{m}}=0.531(0.051) \times \log _{10} t_{\text {max }}-0.116$ (0.065) | $\underline{0.3891}$ | $\bigcirc 0.0001$ | 0.1756 | 172 |
| $\log _{10} T_{\mathrm{m}} \mathrm{f}=0.661$ (0.081) $\times \log _{10} t_{\text {max }}-0.238$ (0.115) | $\underline{0.4785}$ | $\bigcirc 0.0001$ | 0.2212 | 71 |
| $\log _{10} M=-0.816$ (0.045) $\times \log _{10} t_{\text {max }}+0.414$ (0.059) | 0.5980 | $\leq 0.0001$ | 0.2014 | 221 |

All analyses used $\log _{10}$ transformed data. Regression coefficients are shown as $\pm 1 \mathrm{SE}$ in parenthesis; $R^{2}{ }_{\text {adj }}=$ correlation coeficient; $p=$ probablility level; RMSE = root mean square error from regression model, and $n=$ number of stocks (_f after parameter indicates that parameter was estimated for females only). Results for which $p<0.1$ are indicated by underlines and those for which $p<0.05$ by double-underlines.

Table 12. Regression statistics of relationships between biological parameters with $M$ as the independent variable

| Model | $R^{2}{ }_{\text {adj }}$ | $p_{\text {overall }}$ | RMSE | $n$ |
| :---: | :---: | :---: | :---: | :---: |
| $\log _{10} a=-0.266$ (0.126) $\times \log _{10} M-2.301$ (0.087) | $\underline{0.0231}$ | $\underline{0.0361}$ | 0.4168 | 147 |
| $\log _{10} b=0.031(0.008) \times \log _{10} M+0.507$ (0.005) | $\underline{0.0916}$ | 0.0001 | 0.0261 | 148 |
| $\log _{10} L_{\infty}=-0.277(0.074) \times \log _{10} M+1.577$ (0.049) | $\underline{0.0764}$ | $\underline{0.0002}$ | 0.2898 | 158 |
| $\log _{10} \kappa=0.433$ (0.070) $\times \log _{10} M-0.369$ (0.046) | $\underline{0.1913}$ | $\bigcirc$ | 0.2734 | 159 |
| $\log _{10} L_{m}=-0.295(0.155) \times \log _{10} M+1.222$ (0.097) | $\underline{0.0711}$ | $\underline{0.0665}$ | 0.3683 | 34 |
| $\log _{10} L_{\text {m }} \mathrm{f}=-0.110$ (0.064) $\times \log _{10} M+1.491$ (0.047) | $\underline{0.0234}$ | $\underline{0.0891}$ | 0.1796 | 82 |
| $\log _{10} T_{\mathrm{m}}=-0.482(0.047) \times \log _{10} M+0.248$ (0.032) | $\underline{\underline{0.3685}}$ | $\bigcirc$ | 0.1988 | 181 |
| $\log _{10} T_{\mathrm{m}} \mathrm{f}=-0.574(0.092) \times \log _{10} \mathrm{M}+0.272$ (0.070) | $\underline{0.3285}$ | $\bigcirc$ | 0.2438 | 77 |
| $\log _{10} t_{\text {max }}=-0.735$ (0.040) $\times \log _{10} M+0.812$ (0.028) | 0.5980 | $\leq .0001$ | 0.1912 | 221 |

All analyses used $\log _{10}$ transformed data. Regression coefficients are shown as $\pm 1 \mathrm{SE}$ in parenthesis; $R^{2}{ }_{\mathrm{adj}}=$ correlation coeficient; $p=$ probablility level; RMSE = root mean square error from regression model, and $n=$ number of stocks (_f after parameter indicates that parameter was estimated for females only). Results for which $p<0.1$ are indicated by underlines and those for which $p<0.05$ by double-underlines.

Table 13. Multiple regression statistics for relationships between biological parameters and $M$.

| Model | $F$ | $R^{2}$ |
| :--- | :--- | :---: |
| $\log _{10} M=\log _{10} L_{\infty} \times-0.027(0.094)+\log _{10} \kappa \times 0.436(0.093)-0.264(0.138)$ | 18.89 |  |
| $\log _{10} M=\log _{10} t_{\max } \times-0.773(0.073)+\log _{10} L_{\infty} \times 0.014(0.063)+0.327(0.111)$ | 66.96 |  |
| $\log _{10} M=\log _{10} L_{\infty} \times 0.043(0.077)+\log _{10} T_{\mathrm{m}} \times-0.806(0.097)-0.257(0.200)$ | 41.46 | 0.19 |
| $\log _{10} M=\log _{10} t_{\max } x-0.728(0.079)+\log _{10} T_{\mathrm{m}} \times-0.182(0.093)+0.395(0.079)$ | 94.27 | 0.48 |

All analyses used Log ${ }_{10}$ transformed data. Regression coefficients are shown as $\pm 1 \mathrm{SE}$ in parenthesis; $R^{\mathbf{2}}=$ correlation coeficient; $F=\mathrm{F}$ ratio.

Table 14. Comparison of relationships between biological parameters and $M$ derived based on data from the current study and those derived by Hoenig (1983) and Pauly (1980). $R^{2}=$ correlation coeficient; $F=F-$ ratio.

| Model |  | $F$ | $R^{2}$ |
| :---: | :---: | :---: | :---: |
| Hoenig (1983) | $\log _{10} M=1.44-0.982 \times \log _{10} t_{\text {max }}$ |  |  |
| Current study -temperate | $\log _{10} M=1.09-0.858 \times \log _{10} t_{\text {max }}$ | 60.44 | 0.64 |
| Current study -all data | $\log _{10} M=0.97-0.822 \times \log _{10} t_{\text {max }}$ | 327.29 | 0.61 |
| Pauly (1980) | $\log _{10} M=-0.0152-0.2790 \times \log _{10} L_{\infty}+0.6543 \times \log _{10} \kappa+0.4634 \times \log _{10} T$ |  |  |
| Current study -temperate | $\log _{10} M=-0.8803-0.0431 \times \log _{10} L_{\infty}+0.3730 \times \log _{10} \kappa-0.0823 \times \log _{10} T$ | 5.69 | 0.18 |
| Current study -all data | $\log _{10} M=-1.3519+0.0431 \times \log _{10} L_{\infty}+0.4778 \times \log _{10} \kappa+0.2202 \times \log _{10} T$ | 12.54 | 0.27 |

Table 15. Comparison of reproductive parameters among different families of fish. Data for other species from previous studies are included for comparison at the bottom of the table.

| Family | Both sexes |  | Females only |  |
| :---: | :---: | :---: | :---: | :---: |
|  | $T_{\mathrm{m}} / t_{\text {max }}$ | $T_{\mathrm{m}}{ }^{*} M$ | $T_{\mathrm{m}} / t_{\text {max }}$ | $T_{\mathrm{m}}{ }^{*} M$ |
| Acipenseridae |  | 0.65 |  |  |
| Ammodytidae | 0.24 | 1.07 |  |  |
| Anoplopomatidae |  | 0.40 | 0.08 | 0.37 |
| Argentinidae | 0.20 |  |  |  |
| Berycidae |  |  | 0.17 | 0.69 |
| Bothidae |  | 0.76 |  | 0.40 |
| Branchiostegidae |  |  |  |  |
| Carangidae | 0.30 | 1.03 | 0.09 | 0.95 |
| Centrolophidae |  |  | 0.17 | 1.34 |
| Channichthyidae | 0.15 | 1.00 |  |  |
| Cheilodactylida |  |  | 0.07895 | 1.35 |
| Clupeidae | 0.24 |  | 0.20 | 1.00 |
| Cyprinidae |  |  | 0.32 | 0.91 |
| Engraulidae | 0.31 | 1.13 |  |  |
| Esocidae |  | 0.65 |  |  |
| Gadidae | 0.20 | 0.90 | 0.20 | 0.92 |
| Gempylidae | 0.14 | 0.98 | 0.29 | 1.35 |
| Hexagrammidae | 0.27 | 0.40 | 0.24 | 0.96 |
| Lophiidae | 0.29 | 0.60 | 0.29 | 0.60 |
| Lutjanidae | 0.50 | 1.20 | 0.33 | 0.35 |
| Merlucciidae |  |  | 0.28 | 1.40 |
| Moronidae | 0.25 | 0.75 |  |  |
| Mugilidae | 0.19 | 0.99 |  |  |
| Nototheiniidae | 0.44 | 2.03 | 0.45 | 4.98 |
| Nototheniidae | 0.44 | 2.17 | 0.45 | 1.96 |
| Ophidiidae | 0.21 | 1.50 | 0.17 | 0.83 |
| Oreosomatidae |  |  | 0.27 | 1.57 |
| Osmeridae | 0.60 | 3.90 |  |  |
| Pentacerotidae |  | 1.08 |  |  |
| Percidae |  | 1.38 |  |  |
| Platycephalidae |  |  | 0.24 | 1.08 |
| Pleuronectidae | 0.21 | 0.94 | 0.29 | 1.66 |
| Pomatomidae |  |  | 0.09 | 0.28 |
| Sciaenidae |  | 0.30 |  |  |
| Scianidae |  | 0.87 |  |  |
| Scombridae | 0.25 | 1.02 | 0.50 | 1.10 |
| Scophthalmidae |  |  |  | 0.60 |
| Scorpaenidae | 0.11 | 0.76 | 0.15 | 0.64 |
| Sebastidae |  |  | 0.23 | 1.10 |
| Sillaginidae |  |  | 0.25 | 1.80 |
| Soleidae | 0.12 | 0.3 |  |  |
| Sparidae | 0.13 | 0.52 | 0.12 | 0.57 |
| Stromateidae | 0.50 | 1.60 |  |  |
| Synodontidae | 0.43 | 1.24 |  |  |
| Trachichthyidae |  |  | 0.17 | 1.00 |
| Trichiuridae | 0.13 | 0.58 | 0.13 | 0.58 |
| Xiphiidae | 0.56 | 1.00 |  |  |
| Zeidae |  |  | 0.38 | 1.72 |
| Previous studies |  |  |  |  |
| Genus Sebastes |  |  | 0.16 | 0.4 |
| Pacific species |  |  | 0.16 | 0.4 |
| Atlantic species |  |  | 0.23 | 0.7 |


| Clupeidae | 0.24 |  |
| :--- | :--- | :--- |
| Gadiformes | 0.23 | 0.7 |
| Pleuronectiformes | 0.39 | 1.5 |
| Elasmobranchs | 0.38 | 1.6 |
| Birds | 1.7 |  |
| Mammals | 0.4 |  |
| Lizards and snakes | 0.7 |  |



Figure 1. $\mathrm{X}-\mathrm{Y}$ plots of various biological parameters (log-transformed) against the independent variable the logarithm of $a$, with regression line, $95 \%$ confidence intervals and adjusted $R^{2}$ values (* $0.05>P>0.01 ;{ }^{* *} 0.01>P>0.001 ;{ }^{* * *} 0.001>P>0.0001$ ).


Figure 2. X-Y plots of various biological parameters (log-transformed) against the independent variable the logarithm of $b$, with regression line, $95 \%$ confidence intervals and adjusted $R^{2}$ values
$\left(^{*} 0.05>\mathrm{P}>0.01 ;{ }^{* *} 0.01>\mathrm{P}>0.001 ;{ }^{* * *} 0.001>\mathrm{P}>0.0001\right.$ ).


Figure 3. X-Y plots of various biological parameters (log-transformed) against the independent variable the logarithm of $L_{\infty}$, with regression line, $95 \%$ confidence intervals and adjusted $R^{2}$ values
$\left(^{*} 0.05>\mathrm{P}>0.01 ;{ }^{* *} 0.01>\mathrm{P}>0.001 ;{ }^{* * *} 0.001>\mathrm{P}>0.0001\right.$ ).

Information for 'data-rich' species to inform assessments of 'data-poor' species


Figure 4. X-Y plots of various biological parameters (log-transformed) against the independent variable the logarithm of $\kappa$, with regression line, $95 \%$ confidence intervals and adjusted $R^{2}$ values
$\left({ }^{*} 0.05>\mathrm{P}>0.01 ;{ }^{* *} 0.01>\mathrm{P}>0.001 ;{ }^{* * *} 0.001>\mathrm{P}>0.0001\right)$.


Figure 5. X-Y plots of various biological parameters (log-transformed) against the independent variable the logarithm of $L_{m}$, with regression line, $95 \%$ confidence intervals and adjusted $R^{2}$ values ${ }^{*} 0.05>\mathrm{P}>0.01 ;{ }^{* *} 0.01>\mathrm{P}>0.001 ;{ }^{* * *} 0.001>\mathrm{P}>0.0001$ ).


Figure 6. X-Y plots of various biological parameters (log-transformed) against the independent variable the logarithm of female $L m$, with regression line, $95 \%$ confidence intervals and adjusted $R^{2}$ values $\left(^{*} 0.05>\mathrm{P}>0.01 ;{ }^{* *} 0.01>\mathrm{P}>0.001 ;{ }^{* * *} 0.001>\mathrm{P}>0.0001\right.$ ).


Figure 7. $\mathrm{X}-\mathrm{Y}$ plots of various biological parameters (log-transformed) against the independent variable the logarithm of $T_{m}$, with regression line, $95 \%$ confidence intervals and adjusted $R^{2}$ values ${ }^{*} 0.05>P>0.01 ;{ }^{* *} 0.01>P>0.001 ; * * * 0.001>P>0.0001$ ).


Figure 8. X-Y plots of various biological parameters (log-transformed) against the independent variable the logarithm of female $T_{m}$, with regression line, $95 \%$ confidence intervals and adjusted $R^{2}$ values $\left(^{*} 0.05>\mathrm{P}>0.01 ;{ }^{* *} 0.01>\mathrm{P}>0.001 ;{ }^{* * *} 0.001>\mathrm{P}>0.0001\right.$ ).


Figure 9. X-Y plots of various biological parameters (log-transformed) against the independent variable the logarithm of $t_{\text {max }}$, with regression line, $95 \%$ confidence intervals and adjusted $\boldsymbol{R}^{\mathbf{2}}$ values
$\left(^{*} 0.05>\mathrm{P}>0.01 ;{ }^{* *} 0.01>\mathrm{P}>0.001 ;{ }^{* * *} 0.001>\mathrm{P}>0.0001\right.$ ).

Information for 'data-rich' species to inform assessments of 'data-poor' species


Figure 10. X-Y plots of various biological parameters (log-transformed) against the independent variable the logarithm of $M$, with regression line, $95 \%$ confidence intervals and adjusted $R^{2}$ values ( ${ }^{*} 0.05>\mathrm{P}>0.01 ; * * 0.01>\mathrm{P}>0.001 ; * * * 0.001>\mathrm{P}>0.0001$ ).



Figure 11. $\mathrm{X}-\mathrm{Y}$ plot of $\ell \mathrm{n} \tau_{\text {max }}$ against $\ell \mathrm{n} M$, overlayed with the relationship between these variables as estimated by Hoenig (1983) (dark solid line) and the relationships derived using data from this study (light solid line) for all stocks (left panel) and for temperate stocks (right panel).
$\mathbf{9 5 \%}$ confidence intervals about the relationships derived using data from this study are shown as broken grey lines.


Figure 12.X-Y plot of $M$ estimated from the multiple regression based $L_{\infty}, \kappa$ and $T$ for a) all stocks (left panel) and b) temperate stocks (right panel), versus $M$ estimated using Pauly's (1980).

The line indicates the 1:1 relationship.


Figure 13. X-Y plot of $M$ estimated from the multiple regression based $L_{\infty}, \kappa$ and $T$ for all stocks versus $M$ estimated from an equation parameterized by fitting Pauly's (1980) equation to the subset of his data for which $M \leq 0.5 \mathrm{yr}^{-1}$.

The line indicates the 1:1 relationship.

## Chapter 3: Bayesian Analysis of Stock and Recruitment Data

## Introduction

The use of Bayesian techniques when conducting fisheries stock assessments is desirable because inter alia, Bayesian methods provide a single framework within which various sources of uncertainty can be represented (in particular, both parameter and model-structure uncertainty), and because the results from a Bayesian analysis (the probabilities associated with alternative hypotheses) are exactly the information needed when providing scientific management advice to decision makers (Punt and Hilborn, 1997; McAllister and Kirkwood, 1998). However, the primary reason that most stock assessment scientists choose Bayesian over classical approaches is probably because it becomes possible to formally include knowledge from previous assessments (of species / stocks other than that of current interest) in a new assessment. Hilborn and Liermann (1998) argue that using data for well-studied species to inform data-poor species can be considered to be 'standing on the shoulders of giants'.

The frequency with which Bayesian approaches have been used when conducting fisheries stock assessments has increased markedly in recent years (e.g. Hilborn et al., 1994; McAllister et al., 1994; Walters and Ludwig, 1994; Raftery et al., 1995; McAllister and Ianelli, 1997; Punt and Kennedy, 1997; Smith and Punt, 1998; Punt and Butterworth, 1999; Ianelli et al., 2000; Punt et al., 2001a). The main reason for this is the availability of algorithms to numerically evaluate the integrals needed when conducting Bayesian analyses. The two most common methods used to draw samples from the posterior distribution are the Sample-ImportanceResample (SIR) method (Rubin, 1987; Van Dijk et al., 1987) and the Markov-Chain-Monte Carlo (MCMC) algorithm (Hastings, 1970; Gelman et al., 1995).

It is necessary to specify prior distributions for all of the parameters of a model when applying Bayesian techniques (Punt and Hilborn, 1997). Unfortunately, the specification of prior distributions is often the most difficult step of any Bayesian analysis. Priors can either be informative or uninformative. Three types of priors have been used when conducting fisheries stock assessments.

1. Uninformative priors are assigned to parameters for which the data for other species cannot be used to derive an informative prior (e.g. the virgin biomass, $B_{0}$, and fishery catchability).
2. Uniform priors are assigned to parameters for which plausible lower and upper bounds are available (e.g. natural mortality, the parameters that define fishery and survey selectivity).
3. Informative priors are assigned to parameters for which the data for other species can be used to develop a prior for the stock under consideration (e.g. steepness, $h$, survey catchability, and the extent of variation in recruitment, $\sigma_{R}$ ).
This chapter addresses the question of how best to specify priors for $h$ and $\sigma_{R}$ based on data for other stocks /species. Steepness, $h$, is defined as the fraction of virgin recruitment to be expected when the spawning biomass is reduced to $20 \%$ of its virgin level (Francis, 1992). Steepness can be calculated when the relationship between spawning biomass and recruitment is assumed to be governed by the Beverton-Holt or Ricker formulations, i.e.:

$$
\hat{R}= \begin{cases}\frac{\alpha_{1} S}{\beta_{1}+S} & \text { Beverton-Holt }  \tag{3.1}\\ \alpha_{2} S e^{-\beta_{2} S} & \text { Ricker }\end{cases}
$$

where $\hat{R} \quad$ is the model-predicted recruitment,
$S$ is the spawning biomass,
$\alpha_{1}, \beta_{1}$ are the parameters of the Beverton-Holt stock-recruitment relationship, and
$\alpha_{2}, \beta_{2}$ are the parameters of the Ricker stock-recruitment relationship.
Steepness is constrained to lie between 0.2 and 1 for the Beverton-Holt stock-recruitment relationship while steepness must be larger than 0.2 for the Ricker stock-recruitment relationship. The analyses of this chapter place an upper bound of 5 on steepness when the stock-recruitment relationship is assumed to be of the Ricker form. The limit of 5 is imposed to avoid high posterior probability being assigned to stock-recruitment relationships which exhibit severe over-compensation.

Equation (3.1) can be reparametersized in terms of the steepness of the stock-recruitment relationship, the virgin recruitment, $R_{0}$, and the spawner-biomass-per-recruit in the absence of exploitation, $\tilde{S}$, i.e.:

$$
\hat{R}=\left\{\begin{array}{l}
\frac{4 h R_{0} S}{R_{0} \tilde{S}(1-h)+(5 h-1) S}  \tag{3.2}\\
(5 h)^{1.25} S / \tilde{S} e^{-1.25 \ln (5 h) S /\left(\tilde{S} R_{0}\right)}
\end{array}\right.
$$

## Beverton-Holt

Ricker
The value of $S$ is computed from information on natural mortality-at-age and fecundity-at-age using the standard spawning biomass-per-recruit equations. Dorn (2002) notes that an implicit assumption associated with calculating $S$ this way is that all compensation is assumed to occur in the relationship between spawning biomass and subsequent recruitment. There are usually too few data to allow a quantitative evaluation of whether other population dynamics processes (such as growth, natural mortality, fecundity) are, in fact, density-dependent. However, for some species, there is evidence that this is the case (Patterson et al., 2001). For example, in the context of the SESS, the growth of blue grenadier (Macruronus novaezelandiae) off Australia appears to be density-dependent (Punt and Smith, 2001)
At present, when conducting assessments of SESSF species, the values for $h$ and $\sigma_{R}$ are generally prespecified (e.g. Punt et al., 2001a) based on the list of values for $h$ and $\sigma_{R}$ for a range of stocks assembled by McAllister et al. (1994). The estimates of $h$ and $\sigma_{R}$ considered by McAllister et al. (1994) were based on a subset of the data in Myers et al. (1995). Sensitivity is then explored to alternative plausible values. The approach applied by McAllister et al. (1994) is tantamount to applying an Empirical Bayes approach to developing priors for $h$ and $\sigma_{R}$. However, this approach can be criticized because each stock is given equal weight, even though the estimates of $h$ and $\sigma_{R}$ for some of the stocks would be highly imprecise.

Liermann and Hilborn (1997) introduced hierarchical meta-analysis to fisheries assessment by conducting a metaanalysis of the impact of depensation at low stock size. In common with tabling estimates of quantities of interest, hierarchical modeling is a Bayesian technique that can be used to combine data from several independent sources (species / stocks) and represent the outcome in the form of a probability distribution for the quantity of interest. The basic idea is that each species / stock for which data are available has a different value for the quantity of interest but that species / stocks are interchangeable in the sense that the value of the quantity of interest for any given stock can be considered to be a random selection from an underlying distribution (which is the same for all species). These assumptions are displayed graphically in Figure 14 (modified from Liermann and Hilborn (1997)).


Figure 14. Overview of the structure of a hierarchical model illustrating the relationship between the value of some quantity of interest $X$ for a set of stocks, the data for those stocks, and the underlying distribution for $X$.

Hierarchical meta-analysis has been used in the past to examine the steepness of the stock-recruitment relationship (Myers et al., 1999; Dorn, 2002), the relationship between catch rate and abundance (Harley et al., 2001), survey catchability and selectivity (Harley and Myers, 2001; Millar and Methot, 2002), and carrying capacity (Myers et al., 2001).

In order to conduct a hierarchical meta-analysis, it is therefore necessary to:

1. select a set of species / stocks that can be considered be to interchangeable in terms of the quantity (quantities) of interest,
2. select a distribution from which the values for the quantity of interest for each stock is selected,
3. select a model of the data for each stock,
4. specify prior distributions (for the parameters of the prior for the quantity of interest (the hyperprior) and for all of the parameters of the model of the data, except that for the quantity of interest), and
5. sample from the posterior distribution implied by the choices made at steps 2) - 4).

## Methods

## Selection of stocks

Information is available for a total of 443 stocks (Table 1). However, not all of these stocks can be included in the meta-analysis. The reasons for excluding stocks from inclusion in the meta-analysis are.
a) There are fewer than 10 spawning biomass - recruitment pairs ${ }^{1}$.
b) No estimate of spawning biomass-per-recruit is available.

[^1]c) The species is oviviporous - species with this life history strategy appear to have lower steepness than viviparous species (Myers et al., 2002).
Table 16 lists the 128 stocks that remain after the exclusions imposed through constraints a) - c) are applied. The stocks are grouped by order, family, genus and species. Table 16 indicates the number of stocks for each species and for how many stocks of each species is information on each of maximum age, natural mortality rate, age-at-maturity $\left(t_{\mathrm{m}}\right)$, length-at-maturity $\left(l_{\mathrm{m}}\right)$, asymptotic length $\left(\ell_{\infty}\right)$, average latitude, diet, habitat, and depth available.

## Priors and likelihoods

The basic approach taken is virtually identical to that applied by Dorn (2002).

## The model of the data

The available data are estimates of spawning biomass and recruitment. The relationship between these estimates is assumed to be either the Ricker form or the Beverton-Holt form, and the error structure is assumed to be log-normal. For a single species, the contribution of the data to the likelihood function is given by:

$$
\begin{equation*}
L\left(D \mid h, R_{0}, \sigma_{R}\right)=\prod_{i} \frac{1}{\sqrt{2 \pi} \sigma_{R} R_{i}} \exp \left[-\frac{\left(\ln R_{i}-\ln \hat{R}\left(S_{i}\right)+\sigma_{R}^{2} / 2\right)^{2}}{2 \sigma_{R}^{2}}\right] \tag{3.3}
\end{equation*}
$$

where $R_{i} \quad$ is the $\mathrm{i}^{\text {th }}$ recruitment,
$S_{i} \quad$ is the $\mathrm{i}^{\text {th }}$ spawning biomass, and
$\hat{R}(S)$ is the model-predicted recruitment corresponding to a spawning biomass of $S$ (note that $\hat{R}(S)$ depends on $h$ and $R_{0}$ )

Note that, as in Dorn (2002), the likelihood formulation is based on the recruitment from the stockrecruitment relationship being the mean of the distribution rather than the more convention assumption that it is the median of the distribution.
The likelihood of the total (across all species) data set is given by:

$$
\begin{equation*}
L\left(D \mid \underline{h}, \underline{R}_{0}, \underline{\sigma}_{R}\right)=\prod_{k} \prod_{i} \frac{1}{\sqrt{2 \pi} \sigma_{R}^{k} R_{i}^{k}} \exp \left[-\frac{\left(\ell \mathrm{n} R_{i}^{k}-\ell \mathrm{n} \hat{R}^{k}\left(S_{i}^{k}\right)+\left(\sigma_{R}^{k}\right)^{2} / 2\right)^{2}}{2\left(\sigma_{R}^{k}\right)^{2}}\right] \tag{3.4}
\end{equation*}
$$

or

$$
\begin{equation*}
-\ell \mathrm{n} L\left(D \mid \underline{h}, \underline{R}_{0}, \underline{\sigma}_{R}\right) \propto \sum_{k} \sum_{i} \ell \mathrm{n}\left(\sigma_{R}^{k}\right)+\sum_{k} \frac{1}{2\left(\sigma_{R}^{k}\right)^{2}} \sum_{i}\left(\ell \mathrm{n} R_{i}^{k}-\ell \mathrm{n} \hat{R}^{k}\left(S_{i}^{k}\right)+\left(\sigma_{R}^{k}\right)^{2} / 2\right)^{2} \tag{3.5}
\end{equation*}
$$

where $R_{i}^{k} \quad$ is the $\mathrm{i}^{\text {th }}$ recruitment for stock $k$,
$S_{i}^{k} \quad$ is the $\mathrm{i}^{\text {th }}$ spawning biomass for stock $k$,
$\hat{R}^{k}(S)$ is, for stock $k$, the model-predicted recruitment corresponding to a spawning biomass of $S$,
$h^{k} \quad$ is the steepness of the stock-recruitment relationship for stock $k$,
$R_{0}^{k} \quad$ is the virgin recruitment for stock $k$, and
$\sigma_{R}^{k} \quad$ is the standard deviation of the fluctuations about the stock recruitment relationship for stock $k$.

## The prior and hyperprior distributions

The prior for steepness is defined in terms of the logit of steepness. This prior is normal with mean $\mu$ and variance $\tau$, i.e.:

$$
\begin{equation*}
P(\underline{h} \mid \mu, \tau)=\prod_{k} \frac{1}{\sqrt{2 \pi \tau}} \exp \left[-\frac{\left(\tilde{h}^{k}-\mu\right)^{2}}{2 \tau}\right] \tag{3.6}
\end{equation*}
$$

where $\quad \tilde{h}^{k} \quad$ is the logit-transformed steepness, i.e.:

$$
\tilde{h}^{k}= \begin{cases}\ln \left(\frac{h^{k}-0.2}{1-h^{k}}\right) & \text { Beverton-Holt }  \tag{3.7}\\ \ln \left(\frac{h^{k}-0.2}{1-0.2 h^{k}}\right) & \text { Ricker }\end{cases}
$$

The remaining two parameters of the model are virgin recruitment and the standard deviation of the fluctuations about the stock-recruitment relationship, $\sigma_{R}$. The prior distribution for the logarithm of $\sigma_{R}$ is assumed to be uniform over the interval $\mathrm{U}[-\infty, \infty]$ while a relatively uninformative prior is placed on virgin recruitment. This prior, following Dorn (2002), is a normal distribution with mean for stock $k$ given by the average of the observed recruitments for stock $k$ when the observed spawning biomass exceeds the median observed spawning biomass for stock $k$, and a coefficient of variance of 3, i.e.:

$$
\begin{equation*}
P\left(\underline{R}_{0}\right)=\prod_{k} \frac{1}{\sqrt{2 \pi}\left(3 \bar{R}_{0}^{k}\right)} \exp \left[-\frac{\left(R_{0}^{k}-\bar{R}_{0}^{k}\right)^{2}}{2\left(3 \bar{R}_{0}^{k}\right)^{2}}\right] \tag{3.8}
\end{equation*}
$$

where $\bar{R}_{0}^{k} \quad$ is the observed average recruitment when the observed spawning biomass exceeds the median observed spawning biomass.

This prior imposes a weak constraint on the extent to which the virgin recruitment can differ from the observed mean recruitment at 'high' spawning biomass. This prior is not totally ideal because it is based on the data for species $k$. However, in the absence of a prior of this type, the estimate of $h$ can be very close to 0.2 and $R_{0}$ essentially infinite for some stocks.

It is necessary to place a hyperprior on the parameters of the prior for the logit of steepness to finalize the specification of the prior. Following Dorn (2002) again, the hyperprior is chosen to be relatively uninformative so that the posteriors for $h$ are driven primarily by the data rather than by the choice of the prior distribution. In particular, the prior for $\mu$ is assumed to be uniform over a wide interval $[-1000,1000]$ while the prior for $\tau$ is taken to be a scaled inverse chi-squared distribution, i.e.:

$$
\begin{equation*}
P(\tau)=\frac{\left(\frac{v}{2}\right)^{(v / 2)} s^{v} \exp \left(-\frac{v s^{2}}{2 \tau}\right)}{\Gamma\left(\frac{v}{2}\right) \tau^{(v / 2+1)}} \tag{3.9}
\end{equation*}
$$

where $v$ and $s$ are the parameters of the hyperprior.
Gelman et al. (1995) and Dorn (2002) note that this prior is equivalent to basing the prior information for $\tau$ on $v$ observations from $I \mathrm{~V}(\mu, \tau)$ with a mean squared deviation of $s^{2}$. It is necessary to select $v$ and $s^{2}$ to force the prior for $\tau$ away from zero. Dorn (2002) notes that a prior with $v=10$ and $s^{2}=0.5$ is sufficient to achieve this. Given the priors for $h^{k}$ (Equation 3.6), the uniform prior for $\mu$, and the scaled inverse chi-squared prior with $v=10$ and $s^{2}=0.5$ for $\tau$ (Equation 3.9), the prior for $h^{K}$ has high weight at its bounds ( 0.2 and 1 for the BevertonHolt model and 0.2 and 5 for the Ricker model) and is locally flat for values between these bounds.

## Computational aspects

The posterior distribution for the parameters of the hierarchical model is computed using Bayes rule, i.e.:

$$
\begin{equation*}
P\left(\underline{h}, \underline{R}_{0}, \underline{\sigma}_{R}, \mu, \tau \mid D\right)=\frac{L\left(D \mid \underline{h}, \underline{R}_{0}, \underline{\sigma}_{R}\right) P\left(\underline{R}_{0}\right) P(\underline{h} \mid \mu, \tau) P(\tau)}{\iiint L\left(D \mid \underline{h}, \underline{R}_{0}, \underline{\sigma}_{R}\right) P\left(\underline{R}_{0}\right) P(\underline{h} \mid \mu, \tau) P(\tau)} \tag{3.10}
\end{equation*}
$$

The denominator of Equation (3.10) cannot be evaluated numerically so it is necessary to rely on numerical methods to represent the posterior distribution. For the purposes of this study, samples are drawn from the posterior distribution using the Markov Chain Monte Carlo algorithm as implemented in the AD Model Builder package ${ }^{2}$. A total of $20,000,000$ cycles were carried out of which the first $10 \%$ were discarded as a burn-in and the chain was thinned further by sub-sampling every $5,000^{\text {th }}$ element.
A potentially major problem with the use of the Markov Chain Monte Carlo algorithm is how to assess whether it has been run long enough so that convergence to the posterior distribution has been achieved satisfactorily. In this study, this assessment has been achieved in four ways.
a Visually examining the traces for several of the key model outputs.
b Computing the diagnostic statistics developed by Raftery and Lewis (1992), Geweke (1992), and Heidelberger and Welsh (1983).
c Computing the so-called 'single chain Gelman statistic'. This statistic involves comparing the variability of the means in 50 segments of the chain with the variability within each such segment.
d Examining the partial auto-correlation function to assess whether the amount of thinning is sufficient to ensure that sequential points are essentially uncorrelated.

Numerical representations of the posteriors for the parameters of the model (in particular the values for the steepness parameter for each of the 128 stocks) can be constructed from the samples from the posterior

[^2]distribution. A key output from a meta-analysis is the value for the parameter of interest (in this case steepness) for an unknown stock. A numerical representation of the posterior for this quantity is constructed by sampling 100 values from normal distributions defined by each of the $(\mu, \tau)$ pairs in the sample from the posterior distribution and transforming from logit-space to normal space.

## Allowing for covariates

The analysis outlined above is predicated on the assumption that the stocks are interchangeable. However, it may well be that steepness depends on some of the biological characteristics of a stock. For example, Myers et al. (2002) note that steepness depends on the reproductive longevity of a stock. If a relationship between the value of steepness and some biological characteristics is considered plausible, the hyperprior can be extended so that the expected steepness for a stock depends on a vector of covariates that capture these biological characteristics.

In order to assess which (if any) covariates may explain steepness, loess plots of the posterior median for steepness for each stock is plotted against the values for a variety of basic covariates (maximum age, rate of natural mortality, asymptotic length $\left(\ell_{\infty}\right)$, age-at-maturity ( $t_{m}$ ), length-at-maturity ( $l_{\mathrm{m}}$ ), mean latitude, diet, depth and habitat; see Table 16 for a summary of which covariates are available for which stocks). Given the results of Myers et al. (2002), the values for steepness are also plotted against two measures of reproductive longevity (the difference between the maximum age and the age-at-maturity, and the sum of the age-atmaturity and $1 / M$ ).

## Results and discussion

## Evaluation of convergence

Prior to examining the results, it is necessary to evaluate whether the Markov Chain Monte Carlo algorithm has been run for sufficiently long that adequate convergence to the posterior distribution has been achieved. Appendices 3.1 and 3.2 list the posterior medians and $95 \%$ probability intervals for steepness for each stock and for the parameters of the hyperprior $(\mu, \tau)$ and whether convergence has been achieved according to the four convergence statistics. Results are shown in Appendix 3.1 for the two alternative stock-recruitment relationships when the values of the parameters of the hyperprior are assumed to be same for all stocks (abbreviations "Beverton-Holt form" and "Ricker form") and in Appendix 3.2 for the Beverton-Holt form when the values of the parameters of the hyperprior are assumed to be same for all stocks (abbreviation "Beverton-Holt-1") and when the values for these parameters are estimated separately for a group of species consisting of the Clupeiformes, Pleuronectiformes, and Gadiformes and a group consisting of all other species / stocks (i.e. two groups of species) (abbreviation "Beverton-Holt-2").

Figures 15 to 20 show diagnostic plots for $\mu, \tau$, and the model deviance (twice the negative of the logarithm of the likelihood function) based on the fits of the Beverton-Holt-1 and Beveton-Holt-2 analyses. The panels show the trace, the posterior density function (estimated using a normal kernel density), the correlation at different lags, the 50-point moving average against cycle number (dotted line in the rightmost panels), and the running mean and running $95 \%$ probability intervals (solid lines in the rightmost panels). Evidence for lack of convergence would be trends with cycle number in the values for the parameters in the upper left and lower right panels, high auto-correlations in the bottom left panel and a posterior density function which is not smooth.


Figure 15. Diagnostic statistics (see text for details) for the model deviance.


Figure 16.Diagnostic statistics (see text for details) for the mean of the hyperprior for steepness.


Figure 17. Diagnostic statistics (see text for details) for the variance of the hyperprior for steepness.
Considering the results in Figures 15-17 first, there is no evidence for convergence problems for the deviance and for $\mu$. The sample from the posterior for $\tau$ exhibits high lag- 1 autocorrelation. However, $\tau$ does not fail any of the four tests (see Appendix 3.1). The results in Figures 18-20 are more indicative of convergence problems because of the higher auto-correlations and the length of time it takes for the upper and lower 95 percentiles for the posterior to stabilise. This is also evident in Appendix 3.2 where a greater number of parameters fail the tests for lack of convergence. However, visually there are no obvious problems in Figures 15-20.


Figure 18. Diagnostic statistics (see text for details) for the model deviance.





ACF vs. Lag - Chain \#1



Cumulative patterns - Chain \#1


Figure 20. Diagnostic statistics (see text for details) for the variances of the hyperpriors for steepness.

Figures 21-26 summarize the values for six diagnostic statistics (the ratio of the batch standard deviation to the naive standard deviation, the extent of lag-1 auto-correlation, the value of the Raftery-Lewis statistic, the $p$-value computed from the Geweke statistic, whether the Heidelberger and Welch test is passed or not, and the value of the single-chain Gelman statistic) for the estimates of (a) steepness, (b) the logarithm of $B_{0}$ and (c) the logarithm of $\sigma_{R}$ from the Beverton-Holt- 1 and Beverton-Holt-2 analyses. Ideally, the value of the first statistic should be close to 1 , the value of the second statistic should be close to zero, the value of the third statistic should be less than 5 , the value of the fourth statistic should be greater than 0.05 , and the value of the last statistic should be less than 1.05. The $p$-value for the Geweke statistic is less than 0.05 reasonably often (more so for the Beveton-Holt-2 analysis than for the Beverton-Holt-1 analysis). However, this is not a particularly major concern because this statistic can be triggered at random, and the other statistics suggest that convergence has been achieved very successfully.


Figure 21. Summary of six diagnostic statistics for the 128 steepnesses from the Beverton-Holt-1 analysis.


Figure 22. Summary of six diagnostic statistics for the 128 logarithms of $B_{0}$ from the Beverton-Holt-1 analysis.


Information for 'data-rich' species to inform assessments of 'data-poor' species

Figure 23. Summary of six diagnostic statistics for the 128 logarithms of $\sigma_{R}$ from the Beverton-Holt-1

analysis.
Figure 24. Summary of six diagnostic statistics for the 128 steepnesses from the Beverton-Holt-2 analysis.


Figure 25. Summary of six diagnostic statistics for the 128 logarithms of $B_{0}$ from the Beverton-Holt-2 analysis.


Information for 'data-rich' species to inform assessments of 'data-poor' species

Figure 26. Summary of six diagnostic statistics for the 128 logarithms of $\sigma_{R}$ from the Beverton-Holt-2 analysis.

## Ricker or Beverton-Holt stock-recruitment relationship

The modes of the posterior distributions for the steepness of an unknown stock are essentially 1, i.e. recruitment at $20 \%$ of the virgin level equals the recruitment at the virgin level (Figure 27). The posterior for the steepness of the Ricker model assigns high probability to steepness values greater than 1 (Figure 27). A comparison can be made between the Ricker and Beverton-Holt forms for the stock-recruitment relationship using the Deviance Information Criterion (DIC; Spiegelhalter et al., 2002).

$$
\begin{equation*}
D I C=\overline{\text { Deviance }}+p_{D} \tag{3.11}
\end{equation*}
$$

where Deviance is the deviance (averaged over all of the samples from the posterior distribution), and
$p_{D} \quad$ is the 'effective number of parameters'.
The effective number of parameters, $p_{D}$, is defined as the difference between $\overline{\text { Deviance }}$ and the deviance at the maximum of the posterior density function ( Deviance $_{M P D}$ ). Table 17 reports the statistics needed to calculate DIC. Perhaps not expectedly given that a hyperprior is placed on steepness, the effective number of parameters for both forms of the stock-recruitment relationship is notably less than the actual number of parameters (386). The effective number of parameters is greater for the Ricker model than for the BevertonHolt model. This is probably because the hyperprior for steepness included in the Beverton-Holt model constrains the value for steepness more than the hyperprior for steepness included in the Ricker model

because the range for $h$ for the Beverton-Holt model is narrower than that for the Ricker model.
Figure 27. Posterior distributions for the steepness of the stock-recruitment relationship based on the Beverton-Holt (left panel) and Ricker (right panel) stock-recruitment relationship.
DIC clearly favours the Beverton-Holt over the Ricker model; the DIC for the Beverton-Holt model is 888.24 while that for the Ricker model is 1073.67. The very large difference in DIC between these two forms for the stock-recruitment relationship is indicative of the possibility that more complicated models of this relationship (e.g. forms with depensation or the impacts of climate change on productivity) may be supported by the data. However, consideration of such models is beyond the scope of the current project.

Figure 28 shows the posterior distributions for the steepness of an unknown Clupeiform, Pleuronectiform, Gadiform and 'other' stock. The posterior distribution for the first group is essentially identical to that for all stocks in Figure 27. In contrast, the posterior distribution for the steepness for 'other' stocks provides support for lower values of steepness than Figure 27. The difference between the Beverton-Holt-1 and Beverton-Holt2 analyses is explored further in Figure 29 which plots the posterior medians for steepness for the 128 stocks from these two analyses. The points for the 'other' species are frequently smaller for the Beverton-Holt-2 analysis than for the Beverton-Holt-1 analysis. This suggests that the Beverton-Holt-1 analysis 'shrunk' the values of steepness for some of the 'other' species towards those for the clupeiformes, gadiformes and pleuronectiformes. As might be expected, Table 17 indicates that the model in which there are two groups provides a better representation of the data than that which assumes that the steepness for all species / stocks are drawn for a single underlying distribution.


Figure 28. Posterior distributions for the steepness of the stock-recruitment relationship for two species / stock groups based on the Beverton-Holt-2 analysis.


Figure 29. Posterior medians for the steepnesses of the 128 stocks from the analysis that assumes the same hypeprior for all stocks versus the posterior medians for the steepnesses of these stocks from the analysis that assumes different hyperpriors for two groups of stocks.

The posterior for steepness for the clupeiformes, gadiformes and pleuronectiformes can be summarized by its mean, median, and $95 \%$ probability intervals ( $0.866,0.907$, and $[0.606,0.986]$ respectively) while the posterior for the remaining species can be summarized by its mean, median, and $95 \%$ probability intervals ( 0.729 ,
0.757 , and $[0.402,0.959]$ respectively). These posteriors can be summarized adequately by $N\left(-2.00,1.198^{2}\right)$ and $N\left(-0.853,1.216^{2}\right)$ distributions on $\ell n\left(\frac{1-h}{h-0.2}\right)$ (Figure 30).
Appendix 3.3 provides the posterior distributions for Beverton-Holt steepness for each of the 128 stocks considered in the Beverton-Holt-2 analysis. For ease of presentation, two representations of these posteriors are provided: one for the entire $[0.2,1]$ range for steepness and another constrained to the range where the posterior density is non-negligible. The distributions in Appendix 3.3 show that the posteriors for individual stocks (e.g. blue whiting and cod off west Greenland) can differ quite substantially from that for the unknown stock (Figure 28); in particular they can have the bulk of their mass at values for steepness much less than 0.8.

An analysis was conducted in which the clupeiformes, gadiformes and pleuronectiformes were treated as separate groups. However, the results of this analysis are not shown because it failed to exhibit convergence and because it forced the posterior distribution for the steepness for pleuronectiformes to be a delta function at 1 .


Figure 30. Posterior distributions for the logit of ( $h-0.2$ )/0.8 (left panels) and $q-q$ plots of the data in the left panels standardized by their means and variances (right panels).

Results are shown for Clupeiformes, Pleuronectiformes, and Gadiformes in the upper panels and for 'other' species in the lower panels.

## Impact of covariates

$B_{0}$ and $\sigma_{R}$
A key question is whether, as assumed by most Bayesian stock assessments, steepness is independent of other model parameters (in particular the virgin biomass, $B_{0}$, and the extent of variation about the stockrecruitment relationship, $\sigma_{R}$ ). This question is examined in Figure 31 which plots the posterior medians for steepness against those for $B_{0}$ and the posterior medians for $\sigma_{R}$ against those for steepness. There is no evidence in Figure 31 for a relationship between $B_{0}$ and steepness (i.e. there is no evidence that large stocks have higher resilience to exploitation and vice versa). There is a very slight increasing trend in $\sigma_{R}$ with steepness (i.e. more resilient stocks have very slightly higher variation in recruitment). However, overall, there is no evidence that the common assumption made when conducting Bayesian assessments that steepness is independent of $B_{0}$ and $\sigma_{R}$ is violated.


Figure 31. Posterior medians for steepness for the 128 stocks versus the corresponding posterior medians for $B_{0}$ (upper panels) and posterior medians for $\sigma_{R}$ versus the corresponding posterior medians for steepness (lower panels). The right panels plot the $\mathbf{9 0 \%}$ probability intervals for each data point in the left panels.

The mean, median and $90 \%$ intervals of the posterior medians for $\sigma_{R}$ are $0.702,0.615$, and $[0.232,1.525$ ] respectively. Although these summary statistics are not calculated from a distribution for an unknown stock, they nevertheless provide an effective summary of the information about $\sigma_{R}$, and we recommend that account of these values be taken when stock assessments in which recruitment anomalies are estimated are conducted.

Figure 32 summarizes the fits corresponding to the maximum of the posterior density function in terms of the extent of correlation among the residuals about the fit of the stock-recruitment relationship. In general, there is correlation among the residuals, which is primarily positive (mean and median across stocks of 0.307 and 0.326 with $90 \%$ intervals [ $-0.230,0.767]$ ).


Figure 32. Lag-1 correlations among the residuals of the fits of the Beverton-Holt stock-recruitment relationship to the data for the $\mathbf{1 2 8}$ stocks.

The results in this figure are based on the maximum posterior density (MPD) estimates.

## Biological covariates

Figure 33 plots the posterior medians for steepness against reproductive length for two definitions for reproductive length (the age-at-maturity plus $1 / M$ and the difference between the maximum age and the age-at-maturity) and the posterior medians for steepness versus reproductive length (first definition) when the data are divided into categories according to the family (Clupeiformes, Pleuronectiformes, Gadiformes and 'other'), diet preference, depth preference, and habitat preference (see Chapter 2: Species Characteristics for details of the various diet, depth and habitat categories).

There is no obvious relationship between steepness and reproductive length. This is somewhat surprising because Myers et al. (2002) found such a relationship. However, Myers et al. (2002) based their analyses on the Ricker form of the stock-recruitment relationship (which, according to DIC, does not fit the data as well as the Beverton-Holt form) and defined steepness for the Ricker model as $\frac{\alpha_{2}}{4 S \tilde{S} B+\alpha_{2}}$ where $\tilde{S} \tilde{B}$ is the spawner biomass-per-recruit in the absence of exploitation. This is, however, only an approximation to the steepness of the Ricker stock-recruitment relationship (see Equation 3.2) and may be inaccurate for high steepness.

The evidence for different relationships between steepness and reproductive length for different species, habitat preferences, diet preferences, and depth preferences is not clear (Figure 33). The only exceptions to this appear to be that 'other' species with reproductive lengths of $10-15$ years have lower steepnesses, and the stocks with diet type 1 (phytoplankton or zooplankton) have lower steepness at low reproductive length.
Figure 34 plots the posterior medians for steepness against various (continuous) biological characteristics of the stocks under consideration. In general, there is no evidence that steepness depends on any of these characteristics. The results for length-at-maturity should be interpreted with some caution because length-atmaturity is available for only a small subset of the stocks (21 of 128).
There are no major differences in the relationship between steepness and reproductive length among the factors considered. However, steepness does appear to differ among these factors in absolute terms (Figure 35). In particular, and as expected from the results in Figure 28, 'other' species appear to be less resilient than Clupeiformes, Pleuronectiformes, and Gadiformes.


Figure 33. Posterior medians for steepness versus reproductive length for two alternative definitions for reproductive length (upper panels), and steepness versus reproductive length when the data are divided into various discrete categories.

The lines are loess curves.


Figure 34. Posterior medians for steepness versus various biological variables. The solid lines are loess curves.


Figure 35. Box plots of the posterior medians for steepness versus various discrete characteristics.

Figure 36 plots the posterior medians for steepness against various (continuous) biological characteristics of the 'other' stocks. Unfortunately, the number of stocks represented in Figure 36 is low so that it is not straightforward to draw definite conclusions. However, Figure 36 is indicative of lower steepness with higher maximum age and longer reproductive length.


Figure 36. Posterior medians for steepness versus various biological variables for species that are not Clupeiformes, Pleuronectiformes, Gadiformes. The solid lines are loess curves.

## General discussion

Meta-analysis is being used increasingly to improve fisheries stock assessment. For example, in recent years, meta-analysis (based on Bayesian techniques or mixed effects modeling) has been used to examine the steepness of the stock-recruitment relationship (Myers et al., 2002; Dorn, 2002), the maximum rate of increase at low population size (Myers et al., 1999), depensation (Liermann and Hilborn, 1997), trawl survey catchability (Millar and Methot, 2002), survey selectivity (Harley and Myers, 2001), carrying capacity (Myers et al., 2001), and whether catch-rate is likely to be related linearly to abundance (Harley et al., 2001).
Meta-analysis is, however, not without its problems. The major problems as they relate to this study are:

1. Are the stocks truly interchangeable even after the impact of known covariates is removed?
2. Are the data sets representative of the populations to which the results are likely to be applied. This is a particularly important issue for the present study. The stocks for which data on stock and recruitment are available tend to be commercially important species from three genus' (Gadidae, Clupeidae, and Pleuronectidae); three species within these families (Clupea harengus, Gadus morhua, and Melanogrammus aeglefinus) make up 53 of the 128 stocks in Table 16. In contrast, the species to which the results are likely to be applied (Table 18) do not generally come from these families. Furthermore, the stocks in Table 16 tend to be those that are likely to be fairly productive, if only because they have been fished for several decades and are still sufficiently large that quantitative stock assessment methods can be applied to them.
3. Are all the estimates of spawning biomass and recruitment equally reliable? Although some 'quality control' was applied when selecting the 128 species in Table 16, the stock assessments on which the estimates of spawning biomass and recruitment (and spawning biomass-per-recruit) are based are not all likely to be equally reliable.
4. The 'data' are in fact model outputs. One of the implications of this is that the recruitment estimates for the earliest and most recent years are likely to be the least reliable. However, the analysis technique, as currently formulated, gives equal weight to each data point.
5. The results may be influenced by the 'uninformative' priors placed on parameters other than the parameter of interest (in this case the normal prior on $\kappa_{0}$ and the log-uniform prior on $\sigma_{R}$ ), the choice for the form of the prior for steepness (which in this case places high prior weight on very low and very high values for steepness), and the choice for the priors for the parameters of the hyperprior.
Other concerns with meta-analysis which are likely to be less of a concern for this study are:
a 'Publication bias' (studies are only published if the results confirm some apriori hypothesis); stock assessment results are usually reported irrespective of the outcomes.
b Numerical problems conducting the analyses - the application of several different types of diagnostic statistics suggests that this is not likely to be a very major problem in this case.

## Recommendations

1) The data available on stock and recruitment support the Beverton-Holt over the Ricker stockrecruitment relationship, although there are indications that other (more complicated) forms may provide better representations of the existing data.
2) If a single default point estimate for steepness is to be used in a stock assessment based on the Beverton-Holt stock-recruitment relationship, that default should be 0.907 for clupeiformes, gadiformes and pleuronectiformes and 0.757 for 'other' species (the medians of the posteriors for steepness - see Figure 28) while, if a default prior on $\ell \mathrm{n}\left(\frac{1-h}{h-0.2}\right)$ is required, it should be $N\left(-2.00,1.198^{2}\right)$ or $N\left(-0.853,1.216^{2}\right)$ depending on whether or not the species is a clupeiform, gadiform or pleuronectiform.
3) The use of a prior for steepness that is independent of the other model parameters is supported by this study because steepness does not appear to be related to the virgin biomass nor to the extent of variation about the stock-recruitment relationship.
4) The median and $90 \%$ intervals for the posterior medians among stocks for $\sigma_{R}$ are 0.615 and [0.232, 1.525] respectively. These values provide a default value for $\sigma_{R}$ and an appropriate range when conducting tests of sensitivity.

## Table 16.Stocks included in the meta-analysis.

| Species | No <br> Stocks | Max Age | M | $\ell_{\infty}$ | $\boldsymbol{t}_{m}$ | $l_{m}$ | Latitude | Diet | Depth | Habitat |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Aulopiformes |  |  |  |  |  |  |  |  |  |  |
| Synodontidae <br> Harpodon nehereus (bombay duck) <br> Clupeiformes | 1 | 0 | 1 | 1 | 1 | 0 | 1 | 0 | 0 | 0 |
| Clupeidae |  |  |  |  |  |  |  |  |  |  |
| Alosa aestivalis (blueback herring) | 3 | 3 | 3 | 0 | 3 | 0 | 3 | 3 | 1 | 1 |
| Alosa pseudoharengus (anadromous alewife) | 4 | 4 | 4 | 0 | 4 | 0 | 4 | 0 | 0 | 0 |
| Alosa sapidissima (anadromous American shad) | 1 | 0 | 1 | 0 | 1 | 0 | 1 | 0 | 0 | 0 |
| Brevoortia patronus (Gulf menhaden) | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| Brevoortia tyrannus (Atlantic menhaden) | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| Clupea harengus (herring) | 22 | 22 | 19 | 22 | 21 | 1 | 22 | 22 | 22 | 22 |
| Sardina pilchardus (Spanish sardine) | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| Sardinops sagax (sardine) | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| Sprattus sprattus (sprat) | 3 | 3 | 3 | 3 | 3 | 0 | 1 | 1 | 1 | 1 |
| Engraulidae |  |  |  |  |  |  |  |  |  |  |
| Coilia dussumieri (gold-potted grenadier anchovy) | 1 | 0 | 1 | 1 | 1 | 0 | 1 | 1 | 1 | 1 |
| Engraulis encrasicolus (anchovy) | 1 | 1 | 1 | 1 | 1 | 0 | 1 | 1 | 1 | 1 |
| Engraulis mordax (northern anchovy) | 1 | 1 | 1 | 1 | 1 | 0 | 1 | 1 | 1 | 1 |
| Gadiformes |  |  |  |  |  |  |  |  |  |  |
| Gadidae |  |  |  |  |  |  |  |  |  |  |
| Gadus morhua (cod) | 23 | 23 | 23 | 23 | 23 | 3 | 23 | 23 | 23 | 23 |
| Melanogrammus aeglefinus (haddock) | 8 | 8 | 8 | 8 | 8 | 0 | 8 | 8 | 8 | 8 |
| Merlangius merlangus (whiting) | 3 | 3 | 3 | 3 | 3 | 0 | 3 | 3 | 3 | 3 |
| Merluccius bilinearis (silver hake) | 3 | 0 | 3 | 0 | 3 | 0 | 3 | 0 | 0 | 0 |
| Merluccius merluccius (hake) | 1 | 1 | 1 | 1 | 1 | 0 | 1 | 1 | 1 | 1 |
| Merluccius productus (Pacific hake) | 1 | 0 | 1 | 0 | 1 | 0 | 0 | 1 | 1 | 1 |
| Micromesistius poutassou (blue whiting) | 2 | 2 | 2 | 2 | 2 | 1 | 2 | 2 | 2 | 2 |
| Pollachius virens (saithe) | 5 | 5 | 5 | 5 | 5 | 0 | 5 | 5 | 5 | 5 |
| Theragra chalcogramma (walleye pollock) <br> Perciformes | 1 | 1 | 1 | 1 | 1 | 0 | 1 | 1 | 1 | 1 |
| Carangidae |  |  |  |  |  |  |  |  |  |  |
| Trachurus mediterraneus (Mediterranean horse mackerel) | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |



Table 17. Comparison of the Ricker and Beverton-Holt stock-recruitment models using the Deviance Information Criterion.

| Model | $\overline{\text { Deviance }}$ | Deviance $_{\text {MPD }}$ | $p_{D}$ | DIC |
| :---: | :---: | :---: | :---: | :---: |
| Beverton-Holt-1 | 625.58 | 362.93 | 262.65 | 888.24 |
| Beverton-Holt-2 | 632.54 | 388.95 | 243.59 | 876.13 |
| Ricker | 721.60 | 369.52 | 352.08 | 1073.67 |

Results are shown for the Beverton-Holt form for analyses in which the parameters of the hyperprior are assumed to be the same for all stocks (Beverton-Holt-1) and in which these parameters differ among two groups of species (Beverton-Holt-2).

Table 18. The taxonomy of 15 of the 16 SESSF quota species.

| Species | Species |
| :--- | :--- |
| Beryciformes | Gempylidae |
| Berycidae | Rexea solandri (gemfish) |
| Centroberyx affinis (Redfish) | Macruronus novaezelandiae (Blue grenadier) |
| Trachichthyidae | Sillaginidae |
| Hoplostethus atlanticus (orange roughy) | Sillago flindersi (Eastern school whiting) |
| Ophidiiformes | Pleuronectiformes |
| Ophidiidae | Pleuronectidae |
| Genypterus blacodes (ling) | Neoplatycephalus richardsoni (tiger flathead) |
| Perciformes | Sebastinae |
| Carangidae | Sebastidae |
| Pseudocaranx dentex (Silver trevally) | Helicolenus sp. (ocean perch) |
| Centrolophidae | Zeiformes |
| Hyperoglyphe Antarctica (blue-eye trevalla) | Zeidae |
| Seriolella brama (blue warehou) | Zeus faber (John dory) |
| Seriolella punctata (spotted warehou) | Zenopsis nebulosus (Mirror dory) |
| Cheilodactylida |  |
| Nemadactylus macropterus (jackass morwong) |  |

Appendix 3.1. Medians and $95 \%$ probability intervals for steepness and whether four diagnostic tests are failed (indicated by asterisks - 1: Raftery \& Lewis, 2: Geweke, 3: Heidelberger \& Welsh, 4: Single-chain Gelman). The results in this Appendix are based on the Beverton-Holt-1 and Ricker analyses.

| Quantity | Beverton-Holt form |  |  |  |  |  |  | Ricker form |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 2.5\% | Medn | 97.5\% | 1 | 2 | 3 | 4 | 2.5\% | Medn | 97.5\% | 1 | 2 | 3 | 4 |
| Deviance | 573.44 | 624.87 | 680.24 |  |  |  |  | 666.95 | 720.70 | 781.69 |  |  | * |  |
| Hyperprior mean, $\mu$ | 1.589 | 1.925 | 2.287 |  |  |  |  | 0.55 | 0.78 | 1.02 |  |  |  |  |
| Hyperprior variance, $\tau$ | 1.197 | 1.744 | 2.648 |  |  |  |  | 1.02 | 1.30 | 1.96 |  |  |  |  |
| Alosa pseudoharengus (Annaquatucket River, USA) | 0.909 | 0.973 | 0.996 |  |  |  |  | 1.351 | 2.142 | 3.526 |  |  | * |  |
| Alosa pseudoharengus (Damariscotta River, USA) | 0.756 | 0.821 | 0.886 |  |  |  |  | 1.349 | 1.712 | 2.207 |  |  |  |  |
| Alosa pseudoharengus (Lamprey River, USA) | 0.771 | 0.829 | 0.886 |  |  |  |  | 1.189 | 1.497 | 1.965 |  |  |  |  |
| Alosa pseudoharengus (Saint John River, USA) | 0.752 | 0.914 | 0.987 |  |  |  |  | 0.600 | 0.996 | 1.922 |  |  |  |  |
| Engraulis encrasicolus (Black Sea) | 0.448 | 0.498 | 0.581 |  | * |  |  | 0.493 | 0.563 | 0.693 |  |  |  |  |
| Engraulis mordax (California, USA) | 0.531 | 0.904 | 0.993 |  |  |  |  | 0.369 | 0.708 | 1.590 |  |  |  |  |
| Alosa sapidissima (Connecticut River, USA) | 0.745 | 0.896 | 0.990 |  |  |  |  | 1.183 | 1.985 | 3.345 |  |  |  |  |
| Pleurogrammus monopterygius (Eastern Bering Sea / |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Aeultian Islands) | 0.745 | 0.943 | 0.994 |  |  |  |  | 0.800 | 1.764 | 3.635 |  |  |  |  |
| Harpodon nehereus (Northwest coast of India) | 0.392 | 0.415 | 0.446 |  |  |  |  | 0.421 | 0.450 | 0.490 |  |  |  |  |
| Alosa aestivalis (Chowan River, USA) | 0.690 | 0.834 | 0.973 |  |  |  |  | 0.981 | 1.511 | 2.654 |  |  |  |  |
| Alosa aestivalis (Connecticut River, USA) | 0.959 | 0.974 | 0.984 |  |  |  |  | 3.283 | 4.266 | 4.832 |  |  |  |  |
| Alosa aestivalis (Saint John River, USA) | 0.916 | 0.970 | 0.993 |  |  |  |  | 2.015 | 3.237 | 4.498 |  |  |  |  |
| Thunnus obesus (East Pacific) | 0.609 | 0.907 | 0.993 |  |  |  |  | 0.465 | 0.614 | 0.826 |  |  |  |  |
| Pomatomus saltatrix (East Coast, USA) | 0.297 | 0.408 | 0.958 |  |  |  |  | 0.295 | 0.373 | 0.613 |  |  |  |  |
| Thunnus thynnus (West Atlantic) | 0.548 | 0.676 | 0.882 |  |  |  |  | 0.639 | 0.869 | 1.340 |  |  |  |  |
| Micromesistius poutassou (Northern ICES) | 0.697 | 0.933 | 0.993 |  | * |  |  | 0.806 | 1.431 | 2.577 |  |  |  |  |
| Micromesistius poutassou (Southern ICES) | 0.790 | 0.897 | 0.990 |  |  | * |  | 1.517 | 1.847 | 3.522 |  | * | * |  |
| Mallotus villosus (Barents Sea) | 0.875 | 0.963 | 0.995 |  |  |  |  | 1.825 | 3.169 | 4.521 |  |  |  |  |
| Scomber japonicus (Southern California, USA) | 0.522 | 0.903 | 0.992 |  |  |  |  | 0.318 | 0.558 | 1.227 |  |  |  |  |
| Scomber japonicus (Pacific Coast, Japan) | 0.453 | 0.596 | 0.861 |  | * |  |  | 0.472 | 0.670 | 1.132 |  |  |  |  |
| Gadus morhua (West Greenland (NAFO) | 0.674 | 0.853 | 0.969 |  |  |  |  | 0.946 | 1.834 | 3.732 |  |  |  |  |
| Gadus morhua (NAFO 2J3KL) | 0.709 | 0.810 | 0.953 |  |  |  |  | 1.099 | 1.465 | 2.209 |  |  |  |  |
| Gadus morhua (NAFO 3M) | 0.648 | 0.818 | 0.952 |  |  |  |  | 0.952 | 1.873 | 3.747 |  |  |  |  |
| Gadus morhua (NAFO 3NO) | 0.759 | 0.876 | 0.982 |  |  |  |  | 1.327 | 1.918 | 3.390 |  |  |  |  |
| Gadus morhua (NAFO 3Pn4RS) | 0.592 | 0.714 | 0.879 |  |  |  |  | 0.758 | 1.090 | 1.995 |  |  |  |  |


| Quantity |  |  | Beverton-Holt form |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |


| Quantity |  |  | Beverton-Holt form |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |


|  | Quantity | Beverton-Holt form |  |  |  |  |  |  | Ricker form |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | 2.5\% | Medn | 97.5\% | 1 | 2 | 3 | 4 | 2.5\% | Medn | 97.5\% | 1 | 2 | 3 | 4 |
|  | Pleuronectes platessa (ICES IIIa) | 0.713 | 0.939 | 0.994 |  |  |  |  | 1.231 | 2.806 | 4.337 | * |  |  |  |
|  | Pleuronectes platessa (Irish Sea) | 0.897 | 0.974 | 0.997 |  |  |  |  | 2.031 | 2.790 | 3.747 |  |  |  |  |
|  | Pleuronectes platessa (Kattegat) | $0.615$ | $0.749$ | 0.952 |  |  |  |  | 0.815 | 1.179 | 2.206 |  |  |  |  |
|  | Pleuronectes platessa (North Sea) | 0.730 | $0.938$ | 0.995 |  |  |  |  | 1.455 | 2.757 | 4.288 |  |  |  |  |
|  | Pleuronectes platessa (Skagerrak) | 0.661 | 0.920 | 0.993 |  |  |  |  | 0.807 | 1.798 | 3.777 |  |  |  |  |
|  | Pollachius virens (Faroe) | 0.708 | 0.930 | 0.994 |  |  |  |  | 1.171 | 2.509 | 4.249 |  |  |  |  |
| $\underset{\sim}{7}$ | Pollachius virens (Iceland) | 0.803 | 0.951 | 0.995 |  |  |  |  | 1.450 | 2.167 | 3.304 |  |  |  |  |
|  | Pollachius virens (North East Arctic) | 0.821 | 0.881 | 0.941 |  |  |  |  | 1.848 | 2.318 | 2.943 |  |  |  |  |
| $\begin{aligned} & 3 \\ & 0 \\ & \end{aligned}$ | Pollachius virens (North Sea) | 0.795 | 0.887 | 0.978 |  |  |  |  | 1.503 | 2.016 | 2.777 |  |  |  |  |
| 등 | Pollachius virens (ICES VI) | 0.849 | $0.941$ | $0.992$ |  |  |  |  | 1.554 | $1.949$ | $2.456$ |  |  |  |  |
| O | Anoplopoma fimbria (West Coast USA) | $0.454$ | $0.887$ | 0.992 |  |  |  |  | 0.376 | 1.169 | $3.775$ |  |  |  |  |
|  | Sardinops sagax (California) | 0.365 | 0.425 | 0.507 |  |  |  |  | 0.382 | 0.462 | 0.585 |  |  |  |  |
| $\stackrel{\rightharpoonup}{0}$ | Sardina pilchardus (West Iberian (ICES V) | 0.514 | 0.902 | 0.992 |  |  |  |  | 0.503 | 1.702 | 4.075 | * | * | * |  |
| $$ | Thunnus maccoyii (Southern Pacific) | $0.422$ | $0.492$ | $0.587$ |  | * |  |  | $0.447$ | $0.522$ | $0.613$ |  | * |  |  |
| $\stackrel{\rightharpoonup}{\mathrm{n}}$ | Merluccius bilinearis (NAFO 4VWX) | $0.489$ | 0.869 | 0.990 | * |  |  |  | 0.476 | 0.827 | 2.079 |  |  |  |  |
| 0 | Merluccius bilinearis (NAFO 5Ze) | 0.313 | 0.394 | 0.536 |  |  |  |  | 0.315 | 0.405 | 0.572 |  |  |  |  |
| $\stackrel{2}{2}$. | Merluccius bilinearis (Mid Atlantic Bight) | 0.431 | 0.542 | 0.716 |  |  |  |  | 0.454 | 0.605 | 0.890 |  |  |  |  |
| $\stackrel{\rightharpoonup}{\infty}$ | Solea vulgaris (Celtic Sea) | $0.747$ | $0.940$ | $0.995$ |  |  |  |  | $1.248$ | $2.271$ | $3.832$ | * | * |  |  |
| ס | Solea vulgaris (Irish Sea) | $0.683$ | 0.930 | 0.994 |  |  |  |  | 0.989 | 2.850 | 4.517 |  |  |  |  |
| $\stackrel{E}{5}$ | Solea vulgaris (North Sea) | 0.751 | 0.936 | 0.994 |  |  |  |  | 1.050 | 1.855 | 3.221 |  |  |  |  |
| 3 | Solea vulgaris (ICES VIId) | 0.735 | 0.930 | 0.994 |  |  |  |  | 1.253 | 1.814 | 2.772 |  |  |  |  |
| $0$ | Solea vulgaris (ICES VIIe) | $0.510$ | $0.634$ | $0.933$ |  |  |  |  | $0.569$ | $0.716$ | $1.033$ |  |  |  |  |
| $\underset{\sim}{\infty}$ | Sprattus sprattus (Baltic Areas 22-32) | $0.797$ | 0.954 | 0.995 |  | * |  |  | 1.368 | 2.802 | 4.379 |  | * |  |  |
| 0 | Sprattus sprattus (Baltic Areas 26 and 28) | 0.643 | 0.921 | 0.993 |  |  |  |  | 0.548 | 1.143 | 2.859 |  |  |  |  |
| $\stackrel{0}{2}$ | Sprattus sprattus (Black Sea) | 0.428 | 0.579 | 0.818 |  |  |  |  | 0.420 | 0.572 | 0.840 |  |  |  |  |
| $\begin{aligned} & \text { Ғ } \\ & 0 \end{aligned}$ | Xiphias gladius (North Atlantic) | $0.824$ | $0.965$ | $0.996$ | * |  |  |  | 2.872 | 3.533 | 4.131 |  |  |  |  |
| $0$ | Argyrosomus argentatus (East China Sea) | 0.812 | 0.940 | 0.993 |  |  |  |  | 1.520 | 2.419 | 3.793 | * |  | * |  |
| $\stackrel{\Omega}{\sim}$ | Merlangius merlangus (Eastern Black Sea) | 0.458 | 0.883 | 0.991 |  |  | * |  | 0.269 | 0.472 | 1.145 |  |  |  |  |
| - | Merlangius merlangus (Western Black Sea) | $0.418$ | $0.886$ | $0.991$ |  |  |  |  | 0.281 | 0.664 | 2.218 |  |  | * |  |
| $\bigcirc$ | Merlangius merlangus (North Sea) | 0.624 | 0.864 | 0.987 |  |  |  |  | 0.888 | 2.070 | 4.077 |  |  |  |  |
| ${ }^{-}$ | Merlangius merlangus (ICES VIa) | 0.653 | 0.920 | 0.993 |  | * |  |  | 0.796 | 1.817 | 3.658 |  | * |  |  |
| 0180 | Merlangius merlangus (ICES VIId) | 0.573 | 0.837 | 0.989 |  |  |  |  | 0.640 | 1.009 | 2.297 |  |  |  |  |
| $\stackrel{\sim}{\circ}$ | Theragra chalcogramma (Gulf of Alaska) | 0.537 | 0.903 | 0.992 |  |  | * |  | 0.416 | 0.866 | 2.213 |  |  |  |  |


| Quantity | Beverton-Holt form |  |  |  |  |  |  | Ricker form |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 2.5\% | Medn | 97.5\% | 1 | 2 | 3 | 4 | 2.5\% | Medn | 97.5\% | 1 | 2 | 3 | 4 |
| Pleuronectes ferrugineus (NAFO 5Z) | 0.649 | 0.840 | 0.984 |  |  |  |  | 0.826 | 1.304 | 2.641 |  |  |  |  |
| Pleuronectes ferrugineus (Southern New England) | 0.852 | 0.958 | 0.995 |  |  |  |  | 1.752 | 3.120 | 4.464 |  |  |  |  |
| Thunnus albacares (Eastern Pacific Ocea) | 0.709 | 0.923 | 0.993 |  |  |  |  | 0.733 | 1.140 | 1.781 |  |  |  |  |
| Ophiodon elongatus (US West Coast (South) | 0.625 | 0.902 | 0.991 |  |  |  |  | 0.707 | 1.436 | 3.700 |  |  |  |  |
| Ophiodon elongatus (US West Coast (North) | 0.394 | 0.859 | 0.990 |  |  |  |  | 0.332 | 0.572 | 1.617 |  |  |  |  |
| Merluccius productus (US West Coast) | 0.440 | 0.889 | 0.992 |  |  |  |  | 0.300 | 0.795 | 3.037 |  | * |  |  |

Appendix 3.2. Medians and 95\% probability intervals for steepness and whether four diagnostic tests are failed (indicated by asterisks - 1: Raftery \& Lewis, 2: Geweke, 3: Heidelberger \& Welsh, 4: Single-chain Gelman). The results in this Appendix are based on the Beverton-Holt-1 and Beverton-Holt-2 analyses.

| Quantity | Beverton-Holt-1 |  |  |  |  |  |  | Beveton-Holt-2 |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 2.5\% | Medn | 97.5\% | 1 | 2 | 3 | 4 | 2.5\% | Medn | 97.5\% | 1 | 2 | 3 | 4 |
| Deviance | 573.44 | 624.87 | 680.24 |  |  |  |  | 581.399 | 631.764 | 686.873 |  |  |  |  |
| Hyperprior mean, $\mu$ (group-1) | 1.589 | 1.925 | 2.287 |  |  |  |  | 1.702 | 2.015 | 2.382 |  |  |  |  |
| Hyperprior variance, $\tau$ (group-1) | 1.197 | 1.744 | 2.648 |  |  |  |  | 0.166 | 0.822 | 1.711 |  |  | * |  |
| Hyperprior mean, $\mu$ (group-2) |  |  |  |  |  |  |  | 0.938 | 1.421 | 2.244 |  | * |  |  |
| Hyperprior variance, $\tau$ (group-2) |  |  |  |  |  |  |  | 0.548 | 1.223 | 2.788 |  | * | * |  |
| Alosa pseudoharengus (Annaquatucket River, USA) | 0.909 | 0.973 | 0.996 |  |  |  |  | 0.906 | 0.971 | 0.994 |  |  |  |  |
| Alosa pseudoharengus (Damariscotta River, USA) | 0.756 | 0.821 | 0.886 |  |  |  |  | 0.757 | 0.822 | 0.887 |  |  |  |  |
| Alosa pseudoharengus (Lamprey River, USA) | 0.771 | 0.829 | 0.886 |  |  |  |  | 0.773 | 0.831 | 0.889 |  |  |  |  |
| Alosa pseudoharengus (Saint John River, USA) | 0.752 | 0.914 | 0.987 |  |  |  |  | 0.762 | 0.916 | 0.988 |  |  |  |  |
| Engraulis encrasicolus (Black Sea) | 0.448 | 0.498 | 0.581 |  | * |  |  | 0.449 | 0.499 | 0.592 |  |  |  |  |
| Engraulis mordax (California, USA) | 0.531 | 0.904 | 0.993 |  |  |  |  | 0.571 | 0.909 | 0.991 |  |  |  |  |
| Alosa sapidissima (Connecticut River, USA) | 0.745 | 0.896 | 0.990 |  |  |  |  | 0.751 | 0.898 | 0.988 |  |  |  |  |
| Pleurogrammus monopterygius (Eastern Bering Sea / |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Aeultian Islands) | 0.745 | 0.943 | 0.994 |  |  |  |  | 0.600 | 0.876 | 0.984 |  |  |  |  |
| Harpodon nehereus (Northwest coast of India) | 0.392 | 0.415 | 0.446 |  |  |  |  | 0.393 | 0.414 | 0.449 |  |  |  |  |
| Alosa aestivalis (Chowan River, USA) | 0.690 | 0.834 | 0.973 |  |  |  |  | 0.692 | 0.841 | 0.975 |  |  |  |  |
| Alosa aestivalis (Connecticut River, USA) | 0.959 | 0.974 | 0.984 |  |  |  |  | 0.959 | 0.974 | 0.984 |  |  |  |  |
| Alosa aestivalis (Saint John River, USA) | 0.916 | 0.970 | 0.993 |  |  |  |  | 0.911 | 0.968 | 0.992 |  |  |  |  |
| Thunnus obesus (East Pacific) | 0.609 | 0.907 | 0.993 |  |  |  |  | 0.538 | 0.807 | 0.976 |  |  |  |  |
| Pomatomus saltatrix (East Coast, USA) | 0.297 | 0.408 | 0.958 |  |  |  |  | 0.291 | 0.384 | 0.819 |  | * |  |  |
| Thunnus thynnus (West Atlantic) | 0.548 | 0.676 | 0.882 |  |  |  |  | 0.541 | 0.660 | 0.842 |  |  |  |  |
| Micromesistius poutassou (Northern ICES) | 0.697 | 0.933 | 0.993 |  | * |  |  | 0.719 | 0.933 | 0.992 |  |  |  |  |
| Micromesistius poutassou (Southern ICES) | 0.790 | 0.897 | 0.990 |  |  | * |  | 0.791 | 0.893 | 0.990 |  | * | * |  |
| Mallotus villosus (Barents Sea) | 0.875 | 0.963 | 0.995 |  |  |  |  | 0.828 | 0.937 | 0.988 |  |  |  |  |
| Scomber japonicus (Southern California, USA) | 0.522 | 0.903 | 0.992 |  |  |  |  | 0.435 | 0.783 | 0.977 |  |  |  |  |
| Scomber japonicus (Pacific Coast, Japan) | 0.453 | 0.596 | 0.861 |  | * |  |  | 0.450 | 0.584 | 0.790 |  |  |  |  |
| Gadus morhua (West Greenland (NAFO) | 0.674 | 0.853 | 0.969 |  |  |  |  | 0.677 | 0.856 | 0.972 |  |  |  |  |
| Gadus morhua (NAFO 2J3KL) | 0.709 | 0.810 | 0.953 |  |  |  |  | 0.714 | 0.813 | 0.953 |  |  |  |  |
| Gadus morhua (NAFO 3M) | 0.648 | 0.818 | 0.952 |  |  |  |  | 0.657 | 0.824 | 0.952 |  |  |  |  |


| Quantity |  |  | Beverton-Holt-1 |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |


|  | Quantity | Beverton-Holt-1 |  |  |  |  |  |  | Beveton-Holt-2 |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | 2.5\% | Medn | 97.5\% | 1 | 2 | 3 | 4 | 2.5\% | Medn | 97.5\% | 1 | 2 | 3 | 4 |
|  | Merluccius merluccius (ICES VIIIab-d, VIIb k) | 0.633 | 0.731 | 0.978 |  |  | * |  | 0.633 | 0.740 | 0.976 |  | * | * |  |
|  | Clupea harengus (Baltic areas 22 and 24) | 0.840 | 0.962 | 0.995 |  |  |  |  | 0.838 | 0.960 | 0.995 |  |  |  |  |
|  | Clupea harengus (NAFO 4-5) | 0.900 | 0.963 | 0.994 |  |  |  |  | 0.896 | 0.962 | 0.994 |  |  |  |  |
|  | Clupea harengus (Central Coast B.C., Canada) | 0.774 | 0.921 | 0.993 |  |  |  |  | 0.778 | 0.920 | 0.989 |  |  |  |  |
|  | Clupea harengus (Downs stock) | $0.731$ | 0.812 | 0.886 |  |  |  |  | 0.732 | $0.814$ | $0.886$ |  | * |  |  |
|  | Clupea harengus (Eastern Bering Sea) | 0.691 | 0.934 | 0.993 |  |  |  |  | 0.698 | 0.933 | 0.993 |  |  |  |  |
|  | Clupea harengus (Georges Bank) | 0.419 | 0.791 | 0.985 |  |  |  |  | 0.460 | 0.821 | 0.984 |  |  |  |  |
|  | Clupea harengus (Gulf of Finland) | 0.671 | 0.906 | 0.993 |  |  |  |  | 0.667 | 0.908 | 0.991 |  |  |  |  |
|  | Clupea harengus (Gulf of Maine) | $0.861$ | $0.961$ | 0.995 |  |  |  |  | 0.859 | $0.958$ | $0.994$ |  |  |  |  |
|  | Clupea harengus (Gulf of Riga) | 0.578 | 0.873 | 0.991 |  |  | * |  | 0.608 | 0.890 | 0.989 |  | * | * |  |
|  | Clupea harengus (Iceland, Spring-spawning) | 0.439 | 0.589 | 0.793 |  |  |  |  | 0.451 | 0.600 | 0.801 |  |  |  |  |
|  | Clupea harengus (Iceland, Summer-spawning) | 0.635 | 0.756 | 0.902 |  |  |  |  | 0.637 | 0.761 | 0.909 |  |  |  |  |
|  | Clupea harengus (Northern- rish Sea) | $0.806$ | $0.931$ | $0.990$ |  |  |  |  | $0.811$ | $0.931$ | $0.991$ |  | * |  |  |
|  | Clupea harengus (Norway, Spring-spawning) | 0.559 | 0.819 | 0.969 |  | * |  |  | 0.582 | 0.830 | 0.967 |  |  |  |  |
|  | Clupea harengus (North Sea) | 0.863 | 0.905 | 0.942 |  |  |  |  | 0.863 | 0.906 | 0.943 |  | * |  |  |
|  | Clupea harengus (North Srait of Georgia, Canada) | 0.657 | 0.791 | 0.936 |  |  |  |  | 0.671 | 0.795 | 0.934 |  |  |  |  |
|  | Clupea harengus (North West Coast Vancouver Island, |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
|  | Canada) | 0.750 | 0.939 | 0.994 |  |  |  |  | 0.751 | 0.935 | 0.992 |  |  |  |  |
|  | Clupea harengus (Prince Rupert District, USA) | 0.817 | 0.941 | 0.992 |  |  |  |  | 0.822 | 0.942 | 0.992 |  |  |  |  |
|  | Clupea harengus (Queen Charlotte Islands, Canada) | $0.811$ | 0.926 | 0.990 |  |  |  |  | 0.812 | $0.925$ | $0.988$ |  |  |  |  |
|  | Clupea harengus (S.E. Alaska, USA) | $0.486$ | 0.899 | 0.991 |  |  |  |  | 0.533 | 0.906 | $0.990$ |  |  |  |  |
|  | Clupea harengus (Southern Strait of Georgia, Canada) | 0.600 | 0.799 | 0.981 |  |  |  |  | 0.614 | 0.806 | $0.979$ |  |  |  |  |
|  | Clupea harengus (South West Coast Vancouver, Canada) | 0.882 | 0.962 | 0.995 |  |  |  |  | 0.878 | 0.960 | 0.994 |  |  |  |  |
|  | Clupea harengus (ICES VIa (south) and VIIIb, c) | $0.703$ | $0.929$ | 0.993 |  | * |  |  | 0.702 | 0.925 | $0.992$ |  |  |  |  |
|  | Scomber scombrus (NAFO 2 to 6) | 0.557 | 0.873 | 0.991 |  |  |  |  | 0.497 | 0.765 | 0.969 |  | * |  |  |
|  | Trachurus mediterraneus (Black Sea) | 0.610 | 0.870 | 0.990 |  |  |  |  | 0.561 | 0.783 | 0.967 |  |  |  |  |
|  | Scomber scombrus (Western ICES) | 0.526 | 0.888 | 0.992 |  |  |  |  | 0.465 | 0.747 | 0.972 |  |  |  |  |
|  | Brevoortia tyrannus (U.S. Atlantic) | 0.924 | 0.966 | 0.994 |  |  |  |  | 0.924 | 0.966 | $0.992$ |  |  |  |  |
|  | Brevoortia patronus (Gulf of Mexico) | 0.582 | 0.748 | 0.950 |  |  |  |  | 0.594 | 0.763 | 0.958 |  | * |  |  |
|  | Hippoglossus stenolepis (North Pacific) | 0.803 | 0.942 | 0.994 |  |  |  |  | 0.813 | 0.938 | 0.992 |  |  |  |  |
|  | Esox lucius (North Basin, Lake Windermere, USA) | $0.631$ | $0.816$ | $0.982$ |  |  |  |  | 0.616 | 0.764 | 0.950 |  |  |  |  |
|  | Esox lucius (South Basin, Lake Windermere, USA) | 0.483 | 0.654 | 0.951 |  | * |  |  | 0.477 | 0.627 | 0.864 |  |  |  |  |
|  | Pleuronectes platessa (ICES VIIe) | 0.759 | 0.837 | 0.960 |  |  | * |  | 0.764 | 0.841 | 0.962 |  |  |  |  |


| Quantity |  |  | Beverton-Holt-1 |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |


| Quantity | Beverton-Holt-1 |  |  |  |  |  |  | Beveton-Holt-2 |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 2.5\% | Medn | 97.5\% | 1 | 2 | 3 | 4 | 2.5\% | Medn | 97.5\% | 1 | 2 | 3 | 4 |
| Theragra chalcogramma (Gulf of Alaska) | 0.537 | 0.903 | 0.992 |  |  | * |  | 0.577 | 0.909 | 0.991 |  |  |  |  |
| Pleuronectes ferrugineus (NAFO 5Z) | 0.649 | 0.840 | 0.984 |  |  |  |  | 0.658 | 0.845 | 0.983 |  |  |  |  |
| Pleuronectes ferrugineus (Southern New England) | 0.852 | 0.958 | 0.995 |  |  |  |  | 0.847 | 0.959 | 0.994 |  | * |  |  |
| Thunnus albacares (Eastern Pacific Ocea) | 0.709 | 0.923 | 0.993 |  |  |  |  | 0.643 | 0.849 | 0.976 |  |  |  |  |
| Ophiodon elongatus (US West Coast (South) | 0.625 | 0.902 | 0.991 |  |  |  |  | 0.575 | 0.798 | 0.976 |  |  |  |  |
| Ophiodon elongatus (US West Coast (North) | 0.394 | 0.859 | 0.990 |  |  |  |  | 0.350 | 0.682 | 0.970 |  |  |  |  |
| Merluccius productus (US West Coast) | 0.440 | 0.889 | 0.992 |  |  |  |  | 0.496 | 0.900 | 0.991 |  |  |  |  |

Appendix 3.3. Posterior distributions for the steepness of the BevertonHolt stock-recruitment relationship for the 128 stocks.






















## Chapter 4: Data for the Example Application of the Model

## Introduction

The example application of the approach to stock assessment that allows 'data-poor' species to obtain information from 'data-rich' species is based on seven species (blue grenadier, gemfish, pink ling, spotted warehou, mirror dory, king dory, and offshore ocean perch). Gemfish has been divided into two stocks (eastern and western) for assessment and management purposes (Rowling, 1994) so the example application is based on a total of eight stocks.
The data available for assessment purposes for these eight stocks include catches, discard rates, lengthfrequencies and age-compositions by fleet and stock, values for biological parameters, and survey estimates of the spawning biomass of blue grenadier in 1994 and 1995. The following sections outline the details of each source of data.


Figure 37. Map of southern Australia highlighting the various zones considered in the analyses of this chapter.

## Selection of fleets

The process of selecting 'fleets' when conducting assessments involves a balance between selecting a large number of 'fleets' to adequately capture the behaviour of fishers and a small number of 'fleets' to avoid having very few data for each of the fleets. The fleets considered in the example application include one non-trawl fleet and four trawl fleets. The four trawl fleets are based on dividing the fishery into eastern and western sectors (East A, East B, East Tas, zones 10, 20 and 30, and West Tas, Western Zone,
zones 40 and 50 of the SESSF (Figure 37) respectively). This split is needed to separate the fisheries for eastern and western gemfish. The fishery is then further divided into winter (June-August) and nonwinter fisheries in the eastern sector, the winter fishery off western Tasmania, and the remainder of the fishery in the western sector. The split of the fishery in the eastern sector is needed to separate out the fishery that targeted primarily eastern gemfish during their winter spawning migration from the fishery that targeted non-spawning gemfish. This split is needed because the length-composition of the catches during the spawning and non-spawning fisheries differ markedly (Punt et al., 2000). The split of the fishery in the western sector allows the winter fishery for spawning blue grenadier to be treated as a separate fishery. The trawl fleets are assigned numbers as follows:

1) eastern sector; non-winter fishery;
2) eastern sector; winter (June-August) fishery;
3) western sector; all catches in zone 50 and non-winter catches in zone 40; and
4) western sector; winter (June-August) fishery in zone 40.

Table 19 lists the catches (aggregated over the years 1986-2002) for 13 'major stocks' and fleets (values in italics) and the fraction of the catch of each stock / fleet by each of these 13 'major stocks' / fleets. 'Major stocks' / fleets are defined here to be those combinations of stock and fleet for which sufficient data are likely to be available on which to base a catch-rate index of relative abundance. The '27.5' under gemfish in the row for 'blue grenadier / fleet 1 ' implies that $27.5 \%$ of the gemfish catch in the eastern sector in the non-winter period was caught in shots that also caught blue grenadier. There are several large numbers in Table 19. For example:

- King dory is caught predominantly in shots that also caught blue grenadier and pink ling.
- Ocean perch off western Tasmania is caught in winter predominantly in shots that also caught blue grenadier while in the eastern sector, they are caught in shots that also caught pink ling.
- Spotted warehou off western Tasmania is caught in winter predominantly in shots that also caught blue grenadier.
- Blue grenadier is caught predominantly with pink ling in the eastern sector in winter.


## Catches

Information on catches is available from the SEF1 and GNO1 databases (which provide the raw logbook information for the trawl and non-trawl sectors), the SEF2 and SAN2 databases (which contain the vertified landings records for the trawl and non-trawl sectors), and from historical records (for the years prior to 1986 (trawl) and 1997 (non-trawl) - data are available for 1985 in the SEF1 database but these data are incomplete).

## Total annual catches

The total catches for the trawl and non-trawl sectors are based on a variety of data sources. The timeseries of trawl catches by species are constructed as follows:
a Blue grenadier:
1979-83: The catches reported by Smith (1994) multiplied the ratio of the total SEF2 catch over 1993-98 to the total SEF1 catch over 1993-98 (1.05).

1984-2001: The estimated total catch within the SEF area based on the Fishery Assessment Report (198489: Verified catch history; 1990-91: SEF1 data; 1992-2001: SEF2 data).
2002: The SEF2 catch extracted from the SEF2 database.
b Eastern Gemfish
1968-2000: The values agreed to by the Eastern Gemfish Assessment Group (Punt et al., 2000).
2001: $\quad$ The estimated total catch within the SEF area based on the Fishery Assessment Report (SEF2 data).

2002: The 2002 SEF2 catch for all gemfish (195.9t) split between eastern and western gemfish based on the SEF1 catches.

## c Western Gemfish

1979-84: The catches recorded in Victorian logbooks as having been taken from South Australia, western Bass Strait and western Tasmania during 1978-85 rescaled so that the catch for 1985 equals that for 1986.

1985: Assumed equal to the catch for 1986.
1986-91: The SEF1 catches multiplied by the ratio of the SEF2 catch for 1993-98 to the SEF1 catch for 1993-98 (1.44 - this is, as expected, slightly larger than the loss rate due to heading and gutting of 1.3).

1992-2001: The estimated total catch within the SEF area based on the Fishery Assessment Report (SEF2 data).
2002: The 2002 SEF2 catch for all gemfish (195.9t) split between eastern and western gemfish based on the SEF1 catches.
d Pink Ling
1976-83: The catches reported by Tilzey (1994) multiplied the ratio of the total SEF2 catch over 1993-98 to the total SEF1 catch over 1993-98 (1.49).

1984-2001: The estimated total catch within the SEF area based on the Fishery Assessment Report (1984-89: Verified catch history; 1990-91: SEF1 data; 1992-2001: SEF2 data).
2002: The SEF2 catch extracted from the SEF2 database.
e Spotted warehou
1979-85: Linear increase in catch from 0 in 1979 to the 1986 catch in 1986.
1986-2001: The estimated total catch within the SEF area based on the Fishery Assessment Report (1986-89: Verified catch history; 1990-91: SEF1 data; 1992-2001: SEF2 data).
2002: The SEF2 catch extracted from the SEF2 database.
f Mirror Dory
1971-2001: The estimated total catch within the SEF area based on the Fishery Assessment Report (1971-83: NSW State catches; 1984-89: Verified catch history; 1990-91: SEF1 data; 19922001: SEF2 data).

2002: The SEF2 catch extracted from the SEF2 database.

## g Offshore Ocean Perch

1977-2001: The estimated total catch within the SEF area based on the Fishery Assessment Report (1977-83: NSW State catches; 1984-89: Verified catch history; 1990-91: SEF1 data; 19922001: SEF2 data).
2002: The SEF2 catch extracted from the SEF2 database.

## h King dory

1979-84: the catches of king dory recorded in Victorian logbooks as having been taken from South Australia, western Bass Strait and western Tasmania during 1978-85 rescaled so that the catch for 1985 equals that for 1986. The estimates of the annual catch of king dory from 1979-84 was estimated by the sum of the catches recorded as king dory, silver dory and unspecified dory less 12 t . The reason for estimating king dory catches this way is that silver dory is an inshore species with low catches in recent years (an average of 12t over 1986-93) but fairly substantial catches of silver dory are recorded in Victorian logbooks (e.g. 63t in 1981). It is assumed here that some king dory catches were mis-recorded as silver dory and the subtraction of $12 t$ is to account for the 'real' catch of silver dory.

1985: Assumed equal to the catch for 1986.
1986-2002: the SEF1 catches extracted from the SEF1 database (the SEF2 catches for king dory appear unreliable - in some years the reported SEF2 catch is less than $1 \%$ of the SEF1 catch).

The time-series of non-trawl catches by species are constructed as follows:
a Pink ling
1977-96: Thomson et al. (2001).
1997-2002: GN01 logbooks.

## b Spotted warehou

1986-96: Blue Warehou Assessment Group (pers. commn).
1997-2002: GN01 logbooks.

It is necessary to split the trawl catches among each of the four trawl fleets. For the years after 1985, the split is based on the records in the SEF1 database except for blue grenadier. For blue grenadier, the SEF1 catches are reported as processed rather than as whole weight (the SEF2 catches are in whole weight). Therefore, when splitting the total trawl catches of blue grenadier to fleet, the SEF1 catches are first multiplied by conversion factors. These conversion factors are 1.4 for fleets 1-3 (all years), 1.2 for fleet 4 (1986-98), and 1.1 for fleet 4 (1999-2002) (Chesson and Staples, 1995; D.C. Smith, MAFRI, pers. commn). The lower factors for fleet 4 reflect a higher proportion of the catch landed whole (particularly in recent years). The catches prior to 1986 are split to fleets based on the split of the total catch over 1993-98.

Table 20 and Figure 38 list the total catches (trawl and non-trawl) by year while Table 21 lists the trawl catches by fleet and year.


Figure 38. Total catches (trawl - solid lines and non-trawl - dotted lines) by year for the eight stocks considered in the analyses of this report.

## Discard rates

Information on the fraction of the catch (in weight) of each species by each trawl fleet which is discarded annually is available from onboard observers (discards by the non-trawl fleet are assumed to be negligible). Two observer programmes, the SMP (Liggins et al., 1997) and the ISMP (Knuckey et al., 1999) have collected onboard data which can be used to estimate discard rates. The data collected by observers are estimates by shot of the weight retained and the weight discarded. The discard rate is simply the ratio of the weight discarded (summed over all shots by a given fleet in a given year) to the total weight (retained and discarded). The data were validated by excluding any records for which the gear code was not bottom or midwater trawl, and in which the catch did not occur in one of SESSF zones 10, 20, 30, 40 and 50. Discard rates for combinations of fleet and year for which there are not at least 20 data points are ignored when fitting the model. The discard rates used when fitting the model are listed (by year, stock and fleet) in Table 22(a). The precision of the discard rate estimates, expressed as coefficients of variation (Table 22b) were obtained by bootstrapping. For each combination of fleet, year and stock, 1,000 pseudo data sets were generated by resampling shots at random, and with replacement, from the actual observer data for the fleet, year and stock under consideration. The sample size for each pseudo data set is assumed to be the same as the actual sample size.

The discard rate estimates for fleet 4 and for the years 1993-95 tend to be very imprecise. This is due primarily to low sample sizes.

## Catch-rates

Catch-rate data constitute the primary source of information to determine trends in population size for the species in the SESS. However, the catch and effort data need to be standardized to (attempt to) eliminate the impact of factors other than changes in abundance on trends in catch-rates (Gavaris, 1980; Kimura, 1981; Vignaux, 1994). Catch-rate indices were developed for a total of fourteen fleets (two series for one fleet for eastern gemfish, two fleets for blue grenadier, four fleets for spotted warehou, four fleets for pink ling, one fleet for western gemfish, and one fleet for mirror dory - see Table 23). One of the standardized catch-rate series for eastern gemfish was that used by the Eastern Gemfish Assessment Group (EGAG) (Punt et al., 2000) and is not discussed further here. Ideally, catch-rate series should be developed for each combination of species and fleet. However, lack of data for some of these combinations (see, for example, the criteria used to select vessels) precludes this.
The catch-effort standardization exercise involved fitting a linear model (with normal error structure) to relate log-transformed catch-rate data to the factors that influence catch-rates. This approach has been used widely to standardize catch and effort data for SESSF species (e.g. Klaer, 1994, 2004; Punt et al., 2001a; Haddon, 2002a, 2002b). Any shots in which the catch was zero were excluded as were any shots not in SESSF zones 10-50, shots for which the recorded effort is zero, and shots for which depth or zone were missing. Prior to fitting the linear models for each fleet and stock, subsets of the catch and effort data were extracted so that the analysis was based on data for those vessels which are likely to have been targeting the stock concerned. The criteria applied to define the vessels that 'target' each stock are:

- Blue grenadier: catch and effort data for at least three years and a median annual catch (over all years for which catch data are reported) of at least 5 t .
- Spotted warehou: catch and effort data for at least three years and a median annual catch of at least 4 t .
- Pink ling: catch and effort data for at least three years and a median annual catch of at least 4 t .
- Eastern and western gemfish: catch and effort data for at least two years and a median annual catch of at least 2 t .
- Mirror dory: catch and effort data for at least three years and a median annual catch of at least 4 t .

These criteria were applied by fleet (rather than for all fleets combined). Table 24 lists the number of records by fleet and stock for which zone is $10-50$ and the number of records selected for use in the various catch-effort standardizations. The difference between the total number of records for a year and the number actually used relate to records with zero catch or effort, records for which depth is missing and records for vessels that do not satisfy the criteria listed above.


Figure 39. Box and whisker plots examining the relationship between the catch-rate of blue grenadier in zone 50 and zone 40 in September-May (fleet 3) and various potential explanatory variables.

The proportion of zero shots for the vessels selected using the above criteria and for which depth is available ranges from $28 \%$ (blue grenadier in winter in zone 40) to $>70 \%$ (the two summer fisheries for spotted warehou). There are trends in the proportion of zero shots for several of the fleet / species combinations. The most marked of these is for the spawning season fishery for blue grenadier where the proportion of zero shots by the selected vessels declines from $\sim 45 \%$ in $1985-7$ to $15 \%$ in $2000-2$. It is
difficult to interpret changes in the proportion of zero shots as vessels in the SESSF exhibit complex targeting strategies which cannot be resolved from the data collected in logbooks.
The linear models fitted to the catch and effort data include the following factors:

- Year.
- Month.
- Zone.
- Depth: The depth recorded for the trawl assigned to categories based on 100 m depth intervals from 01000 m and categories for $1000-2000 \mathrm{~m}$ and $2000 \mathrm{~m}+$.
- Vessel: An individual factor for each vessel. Vessel was defined by a unique callsign in the SEF1 database.
- Week. Week is defined as the integer part of the Julian day of the year (ignoring leap years) divided by seven. Week cannot be included in models in which month is treated as a factor.
- Covariate species catch. The catch of species (other than that of interest) can be included as a covariate in the analysis. Rather than attempting to develop a functional relationship between the catch-rate of the species of interest and that of each covariate species, the catch of the covariate species is divided into 20 intervals based on the logarithm of the catch (and these intervals form the basis for a categorical variable).

Separate factors are estimated for each vessel rather than characterizing vessels by means of physical characteristics such as length, breadth and horsepower, because these physical factors do not account for differences among vessels in fishing practices (and other factors such as skipper and quota holdings).

The models chosen for each species and fleet are:

- Blue grenadier (abbreviation BG); fleet 3:
- Year+Week+Zone+Depth+Vessel+Pink ling + Jackass morwong + Zone*Week
- Blue grenadier; fleet 4 :
- Year+Week+Depth+Vessel+Pink ling+Jackass morwong
- Spotted warehou (abbreviation SW); fleets 1-3:
- Year+Week+Zone+Depth+Vessel + Pink ling + Jackass morwong + Zone*Week
- Spotted warehou; fleet 4:
- Year+Week+Depth+Vessel+Pink ling+Jackass morwong
- Pink ling (abbreviation PL); fleets 1-3:
- Year+Month+Depth+Vessel+Zone*Vessel+Month*Depth
- Pink ling: fleet 4 :
- Year+Month+Depth+Vessel+ Month*Depth
- Eastern gemfish (abbreviation GF); fleet 2:
- Year+Month+Depth+Vessel+Zone
- Western gemfish; fleet 3:
- Year+Month+Depth+Vessel+Zone
- Mirror dory (abbreviation MD); fleet 2:
- Year+Month+Depth+Vessel+Zone + Eastern gemfish

The choice of factors to include in each model was based on previous analyses by SEFAG (e.g. Haddon, 2002a, 2000b).

It is beyond the scope of this project to conduct a detailed analysis including model selection and regression diagnostics for each of the thirteen catch-effort standardizations. However, diagnostic plots (e.g. Figures 39 and 40) were developed to allow practioners to examine the relationship between catchrate and a variety of potential explanatory variables and whether the specific regression model provides an adequate fit to the data (in the sense of lack of model mis-specification and homoscedascity).
The standardized catch-rate indices used in the analyses are listed in Table 23 and displayed in Figure 41. Figure 41 provides the geometric mean catch rates (dotted lines) as well as the standardized catch-rate indices (solid lines) to allow the impact of standardizing the catch and effort data to be examined. The trend in standardized catch-rate will differ from that of the geometric mean catch-rate if some of the factors in the analysis have changed over time (e.g. changes in the composition of the fleet), while the average absolute value of the standardized catch-rate series may differ from that of the geometric mean catch-rate series depending on the vessel, week, etc. chosen as the standard factors for the analysis. Catch-rates are treated as relative indices of abundance in assessments, so their scale is irrelevant from the viewpoint of the results of an assessment.


Figure 40. Diagnostic plots for the standardization of the catch and effort data for blue grenadier off the west coast of Tasmania (zone 40) in winter (fleet 4).

The solid line in the upper left panel is the standardized catch-rate index and the dotted line is the geometric mean catch rate.


Figure 41. Time-trajectories of standardized catch-rate.
The solid lines are the standardized catch-rate indices and the dotted lines are the geometric mean catch rates.


Figure 42 (continued)Time-trajectories of standardized catch-rate.
The solid lines are the standardized catch-rate indices and the dotted lines are the geometric mean catch rates.

## Length-frequency information

Length frequency data are available from port measurers and from onboard sampling. The former generally involve much larger sample sizes than the latter so the length-frequencies used when fitting the
model and when constructing catch age-compositions are based on the port length measurements only. The onboard sampling programmes provide information on the length-frequencies of the discarded as well as the landed catch and so are used to determine the probability of fish being discarded as a function of size (by fleet and stock).

## Port-based length-frequencies

The data available to construct port-based length-frequencies can be divided into two types: (a) data available at the level of individual landings and (b) data already processed into length-frequencies for the entire catch from a particular area (i.e. cases in which the raw data appear to be no longer available). The sample sizes for the catch length-frequencies are listed in Table 25.

Length data are expressed using a variety of length measurements. Anne Gason and Sonia Talman (MAFRI) were consulted regarding valid length codes, and the equations to convert length measures used to report the catch length-frequencies to the standard length measurement for each species. Data recorded using length measurements that appeared invalid were discarded as were records for which zone was not 10-50 and gear-type was not otter trawl.
Given information on individual landings, the port length-frequencies for a given stock and fleet are constructed from the raw data collected by the measurers using the equation:

$$
\begin{equation*}
N_{y, L}^{s, f}=\sum_{v} \tilde{N}_{y, L}^{s, f, v} / R_{y}^{s, f, v} \tag{4.1}
\end{equation*}
$$

where $N_{y, L}^{s, f}$ is the number of animals in the component of the landed catch of stock $s$ by fleet $f$ during year $y$ that was measured that are in length-class $L$,
$\tilde{N}_{y, L}^{s, f, v}$ is the number of animals in the $v^{\text {th }}$ sample collected from the landed catch of stock $s$ by fleet $f$ during year $y$ that are in length-class $L$, and
$R_{y}^{s, f, v}$ is the fraction of the catch of the $v^{\text {th }}$ sample collected from the landed catch of stock $s$ by fleet $f$ during year $y$ that was measured.
This approach to constructing catch length-frequencies is based on the assumption that the samples for a given fleet and stock are a simple random sample of the catch of that fleet and stock. In principle, this approach to constructing length-frequencies could be generalized so that, for example, port-specific length-frequencies are constructed and these then weighted by the port-specific contribution to the overall catch. The data for the Sydney Fish Market and those collected during the SMP and ISMP are available as individual length measurements.

In the absence of the raw data, Equation 4.1 was not used to construct length-frequencies for:

1) Eastern gemfish (1975-2000) - the length-frequencies collected by port measurers in New South Wales (K. Rowling, pers. commn) were used instead.
2) Blue grenadier in the winter spawning fishery off Tasmania (1999-2002) - the length-frequencies collected by Sarah Russell on trawlers operating in the winter fishery off western Tasmania were used instead.
3) Blue grenadier (1987-89), pink ling (1987-88), king dory (1987-89), western gemfish (1987-88) the length-frequencies collected by port measures at Beachport and Portland were combined to construct these length-frequencies (Smith et al., 1995).
4) Blue grenadier (1984-85) - the length-frequencies were based on sampling off western Tasmania.

The data used when fitting the model were restricted to those combinations of year, fleet and species for which the number of fish measured was at least 200.

## Discard length-frequencies

The proportion of the trawl catch that is discarded by length-class can be determined from the onboard length-frequency data. Equation 4.1 can be used to compute the numbers retained and discarded by length-class, species and fleet. The ratio of the numbers discarded to the numbers discarded and retained empirically defines the probability of discarding by length-class.

The discard length-frequencies were used to determine the values of the parameters of the function that relates the probability of a fish being discarded to length. This function was estimated by fitting a logistic curve to the empirical discard estimates by minimizing the function:

$$
\begin{equation*}
S S=\sum_{f} \sum_{L}\left(P_{L}^{f}-\left[1+\exp \left\{-\ln 19 \frac{\bar{L}_{L}-\phi_{50}}{\phi_{95}-\phi_{50}}\right\}\right]^{-1}\right)^{2} \tag{4.2}
\end{equation*}
$$

where $\bar{L}_{L} \quad$ is the length corresponding to length-class $L$,
$P_{L}^{f} \quad$ is the empirical probability of fleet $f$ discarding a fish of length $L$,
$\phi_{50}$ is the length-at- $50 \%$-retention, and
$\phi_{95}$ is the length-at- $95 \%$ - retention.
The resultant fits are shown in Figure 42. The values for the parameters $\varphi_{50}$ and $\varphi_{95}$ for each stock are listed in Table 26. The fits are generally very good although there are some notable mis-fits (for example, spotted warehou for fleets 3 and 4). These mis-fits suggest that larger (generally marketable) fish are occasionally discarded.

## Age-composition data

Age-length keys (generally sex-specific) are available for six of the seven species. The age-data are generally obtained from samples from the retained component of the catch. Table 27 lists the sample sizes for each stock and year. A total of 303 king dory otoliths have been aged but these data have not been used to construct age-length keys. The age-length keys were provided by the Central Aging Facility, except those for 1980-90 for eastern gemfish, which were provided by K. Rowling (NSW Fisheries).
Validation of the technique used for age determination has been achieved using the bomb radiocarbon method for blue grenadier (Kalish et al. 1997), pink ling (Kalish et al., 2002a), and king dory (Kalish et al., 2002b). Modal progression in the catch age-compostion data has been used to infer that the ageing techniques used for blue grenadier (Punt et al., 2001a), eastern gemfish (this study) and spotted warehou (Thomson, 2003; Taylor and Smith, 2004) are able to determine the primary ages in the catch fairly reliably. Smith et al. (1995) were able to use modal progression to validate the ageing of western gemfish for ages 1-3. The ageing techniques for Ocean perch and mirror dory have not been validated.

Age-composition data (by stock, fleet and year) can be constructed by multiplying the (port) lengthfrequencies by the stock- and year-specific age-length keys (length-at-age is assumed to be independent of fleet). The assumption that length-at-age is independent of fleet will be invalid to some extent if selectivity is strongly size-dependent (Walker et al., 1998) or if there is density-dependent growth (e.g. Punt and Smith, 2001).
There are cases in which length-frequency data exist for some length-classes for which age data are not available. When this happened, the length-classes adjacent to that for which age data were required were investigated and the age data for these length-classes averaged to obtain age data for the length-class for which this was needed. This process of searching adjacent length-classes was repeated if the lengthclasses adjacent to that for which age data were needed also had no age data and this process of an expanding search repeated until ageing data were obtained. Ages greater than 40 were pooled into a plus-group at age 40.

Information for 'data-rich' species to inform assessments of 'data-poor' species


Figure 42. Fits of the logistic curve to the fraction of the catch discarded by fleet and stock.
The dotted line indicates the sample size for each length-class (relative to the sample size for the length-class with the largest sample).



Figure 42 (Continued). Fits of the logistic curve to the fraction of the catch discarded by fleet and stock.


Figure 42 (continued) Fits of the logistic curve to the fraction of the catch discarded by fleet and stock.

The data included in the analyses of this report are restricted to combinations of year and stock for which the sample size on which the age-length key is based is at least 100 animals.

## Age-reading error

Account needs to be taken when fitting population dynamics models of age-reading error. One reason for this is that age-reading error can lead to strong year-classes being 'smeared' across several year-classes, reducing the apparent extent of variation in recruitment. An ageing error matrix can be estimated which determines the probability that an animal of (true) age $a$ is estimated to be age $a^{\prime}$. The data available to estimate the entries in this matrix are the results of inter-reader calibration experiments (i.e. from the ageestimates for otoliths that were read by more than one reader or twice by the same age-reader; see Table 28 for sample sizes for the stocks considered in the analyses of this report). The model on which the estimation of the entries in the ageing error matrix is based assumes that readers are correct on average and that agereading error is normally distributed about the true age, i.e.:

$$
\begin{equation*}
P\left(a^{\prime} \mid a\right)=\frac{\exp \left[-\left(a-a^{\prime}\right)^{2} /\left(2 S D_{a}^{2}\right)\right]}{\sum_{a^{\prime \prime}} \exp \left[-\left(a-a^{\prime \prime}\right)^{2} /\left(2 S D_{a}^{2}\right)\right]} \tag{4.3}
\end{equation*}
$$

where $P\left(a^{\prime} \mid a\right) \quad$ is the probability that an animal of (true) age $a$ is aged to be $a^{\prime}$, and
$S D_{a} \quad$ is the standard deviation of the age-reading error for animals whose true age is $a$.
The parameters of this model are those which define $S D_{a}$. The coefficient of variation of the age-reading error for ages $1+$ is assumed to be related quadratically to age for the analyses of this report, i.e.:

$$
\begin{equation*}
C V_{a}=C V_{0}+\alpha a+\beta a^{2} \tag{4.4}
\end{equation*}
$$

where $\alpha, \beta$ are estimable parameters, and $C V_{0}$ is the coefficient of variation for the limit $a \rightarrow 0$.
To avoid zero values for $S D_{a}, S D_{0}$ is assumed to be equal to $S D_{1}$.
The data from which the parameters of Equation 4.4 can be estimated are pairs of ages ( $a^{-}, a^{-}$) for each otolith, summarized in the form of triples $\left(a_{i}, a_{i}, n_{i}\right)$ where $n_{i}$ is the number of otoliths for which the first age-reader estimated the age to $a_{i}$ and the second age-reader estimated the age to be $a_{i}$. By definition, the total sample size equals __ . The likelihood function maximized to find the estimates for the parameters of the model is therefore:

$$
\begin{equation*}
L=\prod_{i} \tilde{P}\left(a_{i}^{1}, a_{i}^{2}\right)^{n_{i}} \tag{4.5}
\end{equation*}
$$

where $\tilde{P}\left(a^{1}, a^{2}\right)$ is the probability that the first age-reader ages a randomly sampled animal to be $a^{1}$ when the second age-reader ages it to be $a^{2}$. Assuming that age-reading error for the two age-readers is independent, this probability depends on how age-reading error changes with age (see Equation 4.3) and the frequency of each (true) age-class in the sample, i.e.:

$$
\begin{equation*}
\tilde{P}\left(a^{1}, a^{2}\right) \propto \sum_{a} R(a) P\left(a^{1} \mid a\right) P\left(a^{2} \mid a\right) \tag{4.6}
\end{equation*}
$$

where $R(a)$ is the relative frequency of animals of (true) age $a$ in the sample, and $P\left(a^{1} \mid a\right)$ and $P\left(a^{2} \mid a\right)$ are calculated from Equation 4.3. Equation 4.6 implies, for example, that age-readers may agree on an age for a given animal because: (a) they both age it correctly, (b) they both under-estimate its age by 1 year, (c) they both over-estimate its age by 1 year, etc.
The values for $R(a)$ cannot be determined directly from the sample information (because the true ages for the animals are unknown) and are therefore treated as nuisance parameters that need to be estimated along with $C V_{0}, \alpha$, and $\beta$. Note that if the ages in the data set range from 0 to $n$, it is only necessary to estimate $n$ values for $R(a)$ because $R(a)$ is the relative frequency of age-class $a$ in the sample.

## Simulation evaluation

The approach described above involves estimating $n+3$ parameters from the data. The performance of this estimator as the total sample size is changed is evaluated by simulation in the ideal situation in which its key assumptions: (a) the coefficient of variation of ageing error changes quadratically with age, and (b) the probability of the first age-reader making an error is independent of the second age-reader making an error, are valid. The true age-structure of the sample is assumed to be based on a data set for spotted warehou (Figure 43) - although the estimator is unaware of this - for the purposes of these simulations.


Figure 43. Relative frequency of different ages in the population for the simulations used to evaluate the performance of the estimator of age-reading error.

Figure 44 shows histograms (based on 500 simulations for each sample size) of the estimate of the parameter $\iota v_{0}$ and the coefficient of variation for ages 9 and 18 (the midpoint and upper end of the age-range in Figure 43 ). The true values for these quantities are $0.2,0.1$ and 0.15 (i.e. the ability to estimate age is best for intermediate ages). This is not an unreasonable assumption because it is often difficult to identify the annulus for the first year (which leads to a larger percentage error for young animals) and because the annuli become more difficult to distinguish as the number of annuli increase. As expected, performance improves as a function of sample size. The estimates for a total sample size of 100 are both biased and imprecise. Performance in terms of estimating $\mathcal{L} \nu_{0}$ and the coefficient of variation for age 9 improve dramatically when the sample size is increased from 100 to 1,000. However, the estimate of the coefficient of variation for age 18 remains slightly ( $\sim 5 \%$ ) biased and fairly imprecise. The results for sample sizes of 10,000 and 100,000 are qualitatively identical to those for a sample size of 1,000.


Figure 44. Distributions of the relative error associated with estimates of the coefficients of variation of age-reading error for three ages ( 0,9 and 18).

Results are shown for four sample sizes.
The results of the simulations suggest that the precision of the method outlined above is unlikely to produce particularly accurate and precise estimates for sample sizes of $\sim 100$ but that performance is likely to be adequate for sample sizes of $\sim 1,000$.

## Application to actual data

The upper panels of Figures 45 (a)-(g) show the model-predictions of the coefficients of variation and standard deviations as a function of true age for seven of the eight stocks (the data set for king dory being too small to warrant application of the method). The number of data points for each stock are 3789, 1401, 1114, 401, 372, 217, 119 for blue grenadier, pink ling, spotted warehou, western gemfish, eastern gemfish, ocean perch, and mirror dory respectively.
The quality of the fits is examined in the lower panels of these figures which plot the observed number of animals for each combination of age by the first age-reader and age by the second age-reader with the model prediction based on Equation 4.6 (i.e. $n_{i}$ in Equation 4.5 versus $P\left(a^{1}, a^{\alpha}\right)$ in Equation 4.6). The fits are generally adequate (the observed and model-predicted values lie close to the $1: 1$ line). However, there are two noteworthy exceptions. First, the model for blue grenadier is unable to mimic the ability to correctly age animals of age 6 as age 6 and animals of age 7 as age 7 . The exact reasons for this are unclear but may be related to the fact that two very strong year-classes have passed through the blue grenadier stock recently. The second exception is that the fit for mirror dory appears very imprecise (although the sample size for this stock is the lowest of all those in Figure 45).


Figure 45. Diagnostic plots for the fits of the age-reading error model to the data for the seven of the eight stocks.


Figure 45 (continued) Diagnostic plots for the fits of the age-reading error model to the data for the seven of the eight stocks.


Figure 45 (continued) Diagnostic plots for the fits of the age-reading error model to the data for the seven of the eight stocks.


Figure 45 (continued) Diagnostic plots for the fits of the age-reading error model to the data for the seven of the eight stocks.

Figure 46 show the relationships for each of the seven stocks between the coefficient of variation of agereading error and age, with asymptotic $95 \%$ confidence intervals. Three patterns are evident in Figure 46 a) quadratic with a minimum at intermediate ages (blue grenadier, eastern gemfish, pink ling, and ocean perch), b) linear decreasing (spotted warehou and western gemfish), and c) quadratic with a maximum at intermediate ages (mirror dory). The results for mirror dory should, however, treated with some caution as they are fairly imprecise. The precision of the estimates of the coefficients of variation are greatest for intermediate ages (Figure 46) and lowest at high age (the results for mirror dory being an exception to this). This is, however, not particularly surprising because the number of data points is greatest for the youngest and intermediate ages.


Figure 46. Relationships between the coefficient of variation (with asymptotic $95 \%$ confidence intervals) and age for the seven stocks.

## Survey estimates of spawning biomass

Two estimates of the abundance of the spawning stock of blue grenadier are available based on the annual egg production method (59,080t (CV 0.189) and 44,195t (CV 0.287) for 1994 and 1995 respectively; Bulman et al., 1999). These estimates relate to only the female component of the spawning stock because of uncertainties surrounding the sex ratios of spawning blue grenadier (Smith, 1998).

## Biological parameters

Table 26 lists the values for the pre-specified parameters of the model. For the purposes of the example application, growth is assumed to be governed by the von Bertalanffy growth equation and weight related to length by means of a power relationship, i.e.:

$$
\begin{align*}
& L_{a}^{g, s}=\ell_{\infty}^{g, s}\left(1-e^{-\kappa^{g, s}\left(a-t_{0}^{g, s}\right)}\right) \\
& w_{a}^{g, s}=a^{g, s}\left(L_{a}^{g, s}\right)^{b^{g, s}} \tag{4.7}
\end{align*}
$$

where $\quad \ell_{\infty}^{g, s} \quad$ is the asymptotic size of a fish of stock $s$ and sex $g$,
$\kappa^{g, s} \quad$ is the growth rate parameter for a fish of stock $s$ and sex $g$,
$t_{0}^{g, s} \quad$ is the 'age at zero length' for a fish of stock $s$ and sex $g$, and
$a^{g, s}, b^{g, s}$ are the parameters of the weight-length relationship for stock $s$ and sex $g$.
The variance of the logarithm of length-at-age is assumed to change linearly with length, i.e.

$$
\begin{equation*}
\sigma_{a}^{g, s}=\sqrt{\left(\tilde{\sigma}_{0}^{g, s}\right)^{2}+\left[\left(\tilde{\sigma}_{50}^{g, s}\right)^{2}-\left(\tilde{\sigma}_{0}^{g, s}\right)^{2}\right] \frac{L_{a}^{g, a}}{50}} \tag{4.8}
\end{equation*}
$$

where $\tilde{\sigma}_{0}^{g, s}$ is the standard deviation of the logarithm of the length of a fish of stock $s$ and sex $g$ whose expected length is 0 cm , and
$\tilde{\sigma}_{50}^{g, s} \quad$ is the standard deviation of the logarithm of the length of a fish of stock $s$ and sex $g$ whose expected length is 50 cm .
The values for $\sigma_{0}^{\prime}$ and $\sigma_{50}$ for each stock and sex were calculated by regressing the variance of the logarithm of length-at-age on mean length-at-age, weighting each data point by its sample size (Figure 47). The data for blue grenadier, eastern gemfish, pink ling, spotted warehou, and western gemfish are suggestive of declining variance with age while the data suggest that the variance for remainder of the species is independent of age.

[^3]

Figure 47.Length-at-age by stock, and the fit of a straight line to the relationship between the variance of the logarithm of length-at-age and mean length-at-age (each data point relates to the length data for a single age-class).

The results for males are in the left panels and those for females are in the right panels.


Figure 47. (continued) Length-at-age by stock, and the fit of a straight line to the relationship between the variance of the logarithm of length-at-age and mean length-at-age (each data point relates to the length data for a single age-class).

The results for males are in the left panels and those for females are in the right panels.

For all stocks except eastern gemfish, fecundity-at-age is assumed to be the product of female weight-at-age (at the start of the year) and a knife-edged maturity function, i.e.:

$$
f_{a}^{s}= \begin{cases}0 & \text { if } a<a_{m}^{s}  \tag{4.9}\\ w_{a}^{\mathrm{f}, s} & \text { otherwise }\end{cases}
$$

where $a_{m}^{s} \quad$ is the age-at-maturity for stock $s$.
For eastern gemfish, fecundity-at-age is assumed to be the product of selectivity to the winter (spawning) fishery and begin-year weight-at-age.
The proportion spawning parameter, $\mu^{\circ}$, only impacts the ability to fit the survey-based estimates of spawning biomass and so $\mu^{\circ}$ can be assumed to be 1 without loss of generality for all species except blue grenadier. It is known from data from New Zealand (Livingston et al., 1997) that not all blue grenadier spawn annually. Assessments in New Zealand (e.g. McAllister et al., 1994) are based on the assumption that 77 \% (the average of the range of $68-85 \%$ reported by Livingston et al. (1997)) of the potential spawners spawn each year. The same assumption is made here.

For those species for which assessments exist, the extent of recruitment variation has been set equal to the values assumed in the most recent assessments: blue grenadier (1 - Punt et al., 2001a), eastern gemfish (0.6Smith and Punt, 1998); spotted warehou (0.6 - Thomson, 2002), and pink ling ( 0.6 - Klaer, 2003). The extent of recruitment variability has been set equal to 0.6 , a value close to the median of the posterior medians for $\sigma_{R}$ obtained in Chapter 3 of 0.615 , for the remaining stocks (Table 26).

## Discussion

The eight stocks considered in this chapter differ markedly in terms of the data available for assessment purposes. Blue grenadier can be considered to be 'data-rich' because information is available on catches from close to the start of the fishery, extensive age- and length-frequency data are available, and estimates of absolute abundance from fishery-independent surveys are available. Ageing of this species also appears reasonably reliable (see Figure 46).
Eastern gemfish, spotted warehou and pink ling can be considered to be 'data-moderate' because indices of abundance based on standardizing catch and effort data are available, and length-frequency and agecomposition data are available for most of the recent years. However, fishery-independent estimates of relative (and absolute) abundance are not available for these species and the catches are subject to substantial uncertainty (particularly for spotted warehou).
The remaining four stocks (western gemfish, mirror dory, king dory and ocean perch) are clearly 'data-poor'. There are no fishery-independent data for these species, catches are uncertain due to substantial discarding, and length-frequency and age-composition information is limited. Of these stocks, the first two have much more data that the last two, making these last two species the most 'data-poor' species considered in this report.

Table 19. Catches (1986-2002; tonnes) by 'major stock' and fleet (values in underlined italics) and the percentage of the catch of the other stocks / fleets caught with these 'major stock' and fleets.

| Species | Fleet | Blue <br> grenadier | Gemfish | Pink ling | Spotted <br> warehou | King dory | Ocean <br> perch | Mirror <br> Dory |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Blue grenadier | 1 | $\underline{7456}$ | 27.5 | 49.6 | 16.0 | 83.1 | 43.2 | 34.2 |
|  | 2 | $\underline{1486}$ | 27.6 | 51.5 | 16.9 | 80.3 | 40.9 | 25.7 |
| Eastern gemfish | 3 | $\underline{16824}$ | 43.0 | 83.5 | 58.4 | 84.8 | 77.6 | 67.5 |
| Pink ling | 4 | $\underline{38227}$ | 54.4 | 86.5 | 91.9 | 84.4 | 90.4 | 78.4 |
|  | 2 | 61.4 | $\underline{13155}$ | 55.5 | 22.0 | 51.3 | 52.5 | 75.4 |
|  | 1 | 76.9 | 44.0 | $\underline{6870}$ | 44.1 | 80.4 | 91.4 | 64.2 |
|  | 2 | 90.3 | 59.6 | $\underline{3816}$ | 47.9 | 86.2 | 88.4 | 69.2 |
| Spotted warehou | 73.4 | 27.8 | $\underline{4977}$ | 55.3 | 82.6 | 79.3 | 53.1 |  |
|  | 3 | 25.0 | 41.2 | $\underline{642}$ | 52.0 | 84.4 | 78.5 | 69.1 |

Table 20. The total annual catches (trawl and non-trawl) by stock.

|  | 1968 | 1969 | 1970 | 1971 | 1972 | 1973 | 1974 | 1975 | 1976 | 1977 | 1978 | 1979 | 1980 | 1981 | 1982 | 1983 | 1984 | 1985 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Trawl catches |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Blue grenadier | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 512 | 857 | 470 | 815 | 941 | 1342 | 1201 |
| Eastern gemfish | 1440 | 1460 | 1470 | 1480 | 1500 | 1460 | 1752 | 1652 | 3392 | 4676 | 6672 | 5162 | 6040 | 4700 | 4230 | 3715 | 3398 | 3406 |
| Ling | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 149 | 224 | 299 | 299 | 448 | 598 | 523 | 672 | 770 | 698 |
| Spotted warehou | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 183 | 365 | 548 | 731 | 914 | 1096 |
| Mirror dory | 0 | 0 | 0 | 40 | 90 | 170 | 200 | 290 | 420 | 580 | 860 | 420 | 290 | 160 | 310 | 220 | 241 | 377 |
| Ocean perch (offshore) | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 200 | 250 | 200 | 200 | 250 | 275 | 400 | 287 | 271 |
| King dory | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 17 | 163 | 147 | 258 | 159 | 133 | 81 | 55 |
| Western gemfish | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 195 | 401 | 492 | 974 | 928 | 757 | 429 | 445 |
| Non-trawl catches |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Ling | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 11 | 54 |
| Spotted warehou | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |


|  | 1986 | 1987 | 1988 | 1989 | 1990 | 1991 | 1992 | 1993 | 1994 | 1995 | 1996 | 1997 | 1998 | 1999 | 2000 | 2001 | 2002 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Trawl catches |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Blue grenadier | 1476 | 2137 | 2197 | 1996 | 2363 | 3690 | 3294 | 3373 | 3172 | 2767 | 3054 | 4548 | 5733 | 9313 | 8655 | 9124 | 9156 |
| Eastern gemfish | 3951 | 4762 | 3745 | 2167 | 1645 | 1067 | 1060 | 1348 | 513 | 512 | 554 | 594 | 391 | 278 | 119 | 106 | 64 |
| Ling | 715 | 882 | 772 | 833 | 730 | 798 | 895 | 1617 | 1434 | 1700 | 1987 | 1987 | 1702 | 1705 | 1605 | 1356 | 1071 |
| Spotted warehou | 1279 | 1147 | 1618 | 1038 | 1359 | 1480 | 818 | 2217 | 2724 | 2535 | 2738 | 2785 | 2410 | 2738 | 3726 | 3294 | 4090 |
| Mirror dory | 487 | 569 | 483 | 650 | 300 | 246 | 214 | 335 | 322 | 281 | 403 | 546 | 425 | 352 | 208 | 303 | 306 |
| Ocean perch (offshore) | 243 | 231 | 256 | 248 | 199 | 249 | 234 | 395 | 355 | 378 | 395 | 464 | 377 | 399 | 381 | 394 | 522 |
| King dory | 55 | 64 | 57 | 99 | 67 | 141 | 112 | 148 | 166 | 168 | 155 | 147 | 123 | 93 | 97 | 141 | 155 |
| Western gemfish | 445 | 361 | 332 | 226 | 196 | 392 | 126 | 135 | 138 | 124 | 208 | 227 | 185 | 272 | 349 | 253 | 132 |
| Non-trawl catches |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Ling | 86 | 88 | 103 | 115 | 82 | 82 | 274 | 615 | 496 | 415 | 591 | 225 | 178 | 247 | 226 | 316 | 465 |
| Spotted warehou | 1 | 1 | 6 | 21 | 24 | 55 | 85 | 61 | 79 | 97 | 165 | 169 | 66 | 31 | 4 | 1 | 1 |

Table 21. Trawl catches ( $\mathbf{t}$ ) by stock and fleet
(a) 1968-1985

| Stock | Fleet | 1968 | 1969 | 1970 | 1971 | 1972 | 1973 | 1974 | 1975 | 1976 | 1977 | 1978 | 1979 | 1980 | 1981 | 1982 | 1983 | 1984 | 1985 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Blue grenadier | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 70 | 117 | 64 | 111 | 128 | 183 | 163 |
|  | 2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 9 | 16 | 9 | 15 | 17 | 25 | 22 |
|  | 3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 170 | 285 | 156 | 271 | 312 | 445 | 399 |
|  | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 263 | 440 | 242 | 419 | 483 | 689 | 617 |
| Eastern gemfish | 1 | 0 | 20 | 30 | 40 | 60 | 20 | 20 | 40 | 40 | 170 | 450 | 460 | 440 | 190 | 270 | 305 | 300 | 205 |
|  | 2 | 1440 | 1440 | 1440 | 1440 | 1440 | 1440 | 1732 | 1612 | 3352 | 4506 | 6222 | 4702 | 5600 | 4510 | 3960 | 3410 | 3098 | 3201 |
| Ling | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 64 | 96 | 128 | 128 | 191 | 255 | 223 | 287 | 329 | 298 |
|  | 2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 31 | 47 | 63 | 63 | 94 | 125 | 109 | 141 | 161 | 146 |
|  | 3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 48 | 72 | 96 | 96 | 145 | 193 | 169 | 217 | 248 | 225 |
|  | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 6 | 9 | 12 | 12 | 18 | 25 | 21 | 28 | 32 | 29 |
| Spotted warehou | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 50 | 101 | 151 | 202 | 252 | 303 |
|  | 2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 46 | 93 | 139 | 185 | 232 | 278 |
|  | 3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 76 | 153 | 229 | 305 | 382 | 458 |
|  | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 10 | 19 | 29 | 38 | 48 | 57 |
| Mirror dory | 1 | 0 | 0 | 0 | 19 | 43 | 80 | 95 | 137 | 199 | 274 | 407 | 199 | 137 | 76 | 147 | 104 | 114 | 178 |
|  | 2 | 0 | 0 | 0 | 12 | 26 | 49 | 58 | 84 | 122 | 168 | 250 | 122 | 84 | 46 | 90 | 64 | 70 | 109 |
|  | 3 | 0 | 0 | 0 | 9 | 20 | 39 | 45 | 66 | 95 | 132 | 195 | 95 | 66 | 36 | 70 | 50 | 55 | 86 |
|  | 4 | 0 | 0 | 0 | 0 | 1 | 2 | 2 | 3 | 4 | 6 | 9 | 4 | 3 | 2 | 3 | 2 | 2 | 4 |
| Ocean perch (offshore) | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 131 | 163 | 131 | 131 | 163 | 180 | 262 | 188 | 177 |
|  | 2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 37 | 46 | 37 | 37 | 46 | 51 | 74 | 53 | 50 |
|  | 3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 28 | 35 | 28 | 28 | 35 | 38 | 56 | 40 | 38 |
|  | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 4 | 5 | 4 | 4 | 5 | 6 | 8 | 6 | 6 |
| King dory | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 2 | 17 | 14 | 42 | 18 | 31 | 15 | 14 |
|  | 2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 3 | 3 | 8 | 3 | 6 | 3 | 3 |
|  | 3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 14 | 154 | 123 | 376 | 157 | 275 | 134 | 121 |
|  | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 13 | 11 | 32 | 14 | 24 | 12 | 10 |
| Western gemfish | 3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 192 | 394 | 483 | 957 | 912 | 744 | 421 | 437 |
|  | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 4 | 7 | 9 | 18 | 17 | 14 | 8 | 8 |

(b) 1986-2002

| Stock | Fleet | 1986 | 1987 | 1988 | 1989 | 1990 | 1991 | 1992 | 1993 | 1994 | 1995 | 1996 | 1997 | 1998 | 1999 | 2000 | 2001 | 2002 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Blue grenadier | 1 | 285 | 504 | 478 | 583 | 724 | 995 | 734 | 792 | 630 | 398 | 330 | 453 | 472 | 994 | 526 | 410 | 206 |
|  | 2 | 99 | 172 | 220 | 108 | 180 | 85 | 37 | 56 | 133 | 16 | 23 | 60 | 132 | 278 | 121 | 135 | 17 |
|  | 3 | 872 | 775 | 1157 | 1271 | 929 | 1811 | 1671 | 1524 | 1191 | 1150 | 1194 | 891 | 1589 | 1738 | 1738 | 788 | 1377 |
|  | 4 | 221 | 686 | 342 | 33 | 531 | 799 | 852 | 1001 | 1218 | 1204 | 1507 | 3144 | 3540 | 6303 | 6270 | 7791 | 7557 |
| Eastern gemfish | 1 | 130 | 80 | 175 | 175 | 80 | 50 | 30 | 312 | 106 | 50 | 66 | 63 | 74 | 40 | 20 | 39 | 25 |
|  | 2 | 3821 | 4682 | 3570 | 1992 | 1565 | 1017 | 1030 | 1036 | 407 | 462 | 488 | 531 | 317 | 238 | 99 | 67 | 39 |
| Ling | 1 | 305 | 389 | 364 | 369 | 331 | 388 | 451 | 768 | 691 | 741 | 916 | 746 | 645 | 690 | 524 | 455 | 310 |
|  | 2 | 150 | 242 | 275 | 233 | 234 | 143 | 253 | 380 | 335 | 347 | 328 | 432 | 362 | 473 | 426 | 256 | 205 |
|  | 3 | 231 | 211 | 122 | 225 | 151 | 257 | 183 | 440 | 376 | 519 | 625 | 724 | 632 | 463 | 555 | 561 | 487 |
|  | 4 | 29 | 40 | 12 | 6 | 15 | 11 | 8 | 28 | 32 | 93 | 118 | 85 | 63 | 78 | 100 | 84 | 69 |
| Spotted warehou | 1 | 353 | 186 | 251 | 266 | 310 | 309 | 368 | 502 | 806 | 688 | 1043 | 789 | 439 | 205 | 214 | 261 | 403 |
|  | 2 | 325 | 217 | 675 | 127 | 742 | 473 | 204 | 766 | 1019 | 851 | 371 | 431 | 480 | 694 | 586 | 456 | 366 |
|  | 3 | 534 | 598 | 634 | 641 | 256 | 675 | 222 | 884 | 757 | 780 | 1194 | 1480 | 1321 | 1411 | 1770 | 1570 | 2180 |
|  | 4 | 67 | 146 | 58 | 5 | 51 | 22 | 23 | 65 | 143 | 216 | 130 | 85 | 170 | 427 | 1156 | 1007 | 1141 |
| Mirror dory | 1 | 230 | 172 | 161 | 132 | 125 | 115 | 110 | 171 | 224 | 187 | 167 | 172 | 163 | 113 | 97 | 91 | 92 |
|  | 2 | 141 | 376 | 298 | 504 | 165 | 114 | 92 | 143 | 77 | 48 | 94 | 190 | 120 | 165 | 82 | 78 | 45 |
|  | 3 | 111 | 20 | 23 | 11 | 10 | 17 | 12 | 21 | 20 | 41 | 132 | 179 | 140 | 72 | 27 | 129 | 163 |
|  | 4 | 5 | 2 | 2 | 2 | 0 | 0 | 0 | 0 | 0 | 5 | 10 | 6 | 3 | 3 | 2 | 5 | 6 |
| Ocean perch (offshore) | 1 | 159 | 132 | 165 | 153 | 119 | 168 | 156 | 264 | 262 | 236 | 268 | 279 | 234 | 268 | 252 | 258 | 322 |
|  | 2 | 45 | 43 | 76 | 78 | 62 | 41 | 52 | 94 | 58 | 62 | 65 | 91 | 70 | 85 | 75 | 80 | 107 |
|  | 3 | 34 | 53 | 14 | 17 | 15 | 38 | 26 | 37 | 27 | 68 | 51 | 86 | 65 | 39 | 50 | 52 | 84 |
|  | 4 | 5 | 3 | 0 | 0 | 2 | 1 | 0 | 1 | 7 | 12 | 11 | 9 | 8 | 7 | 4 | 4 | 9 |
| King dory | 1 | 5 | 7 | 10 | 12 | 12 | 17 | 14 | 27 | 16 | 10 | 9 | 14 | 8 | 20 | 13 | 20 | 23 |
|  | 2 | 1 | 1 | 10 | 5 | 3 | 1 | 1 | 5 | 8 | 1 | 2 | 0 | 1 | 3 | 2 | 4 | 1 |
|  | 3 | 45 | 49 | 36 | 80 | 49 | 120 | 91 | 110 | 137 | 144 | 124 | 124 | 104 | 64 | 73 | 109 | 121 |
|  | 4 | 4 | 7 | 1 | 2 | 3 | 3 | 6 | 6 | 5 | 14 | 20 | 8 | 9 | 5 | 10 | 9 | 11 |
| Western gemfish | 3 | 437 | 355 | 330 | 225 | 195 | 392 | 126 | 119 | 138 | 124 | 208 | 226 | 185 | 271 | 346 | 249 | 130 |
|  | 4 | 8 | 6 | 2 | 0 | 1 | 0 | 0 | 16 | 0 | 0 | 0 | 1 | 0 | 1 | 3 | 4 | 2 |

Table 22. The discard rates by stock and fleet. ' ${ }^{\prime \prime}$ ' denotes that data are not available for the stock, fleet and year concerned.
(a) Point estimates

|  | Fleet | 1993 | 1994 | 1995 | 1996 | 1997 | 1998 | 1999 | 2000 | 2001 | 2002 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Blue grenadier | 1 | 0.001 | 0.001 | 0.160 | 0.303 | 0.644 | 0.469 | 0.047 | 0.003 | 0.001 | 0.002 |
|  | 2 | - | - | - | - | 0.535 | - | 0.007 | 0.008 | 0.006 | - |
|  | 3 | - | 0.007 | 0.009 | 0.365 | 0.724 | 0.553 | 0.029 | 0.013 | 0.002 | 0.002 |
|  | 4 | - | - | - | - | - | 0.030 | 0.001 | - | - | - |
| Eastern gemfish | 1 | 0.006 | 0.260 | 0.103 | 0.467 | 0.289 | 0.058 | 0.231 | 0.026 | 0.151 | 0.230 |
|  | 2 | 0.878 | - | 0.117 | 0.115 | 0.052 | 0.067 | 0.181 | 0.222 | 0.055 | 0.206 |
| Pink ling | 1 | 0.003 | 0.005 | 0.009 | 0.020 | 0.067 | 0.026 | 0.004 | 0.004 | 0.006 | 0.013 |
|  | 2 | 0.001 | 0.021 | 0.033 | 0.008 | 0.165 | 0.011 | 0.003 | - | 0.001 | 0.001 |
|  | 3 | - | 0.000 | 0.002 | 0.018 | 0.042 | 0.028 | 0.012 | 0.019 | 0.004 | 0.003 |
|  | 4 | - | - | - | - | - | 0.029 | 0.001 | - | - | - |
| Spotted warehou | 1 | 0.010 | 0.009 | 0.231 | 0.013 | 0.010 | 0.016 | 0.002 | 0.011 | 0.167 | 0.362 |
|  | 2 | 0.078 | 0.005 | 0.587 | 0.198 | 0.255 | 0.055 | 0.003 | 0.002 | 0.208 | 0.155 |
|  | 3 | - | 0.024 | 0.079 | 0.313 | 0.383 | 0.678 | 0.045 | 0.017 | 0.192 | 0.047 |
|  | 4 | - | - | - | - | - | 0.176 | 0.101 | - | 0.011 | - |
| Ocean perch | 1 | 0.209 | 0.717 | 0.799 | 0.575 | 0.432 | 0.305 | 0.241 | 0.367 | 0.673 | 0.344 |
|  | 2 | 0.185 | 0.612 | 0.630 | 0.484 | 0.189 | 0.157 | 0.045 | 0.423 | 0.779 | 0.288 |
|  | 3 | - | 0.254 | 0.009 | 0.014 | 0.002 | 0.005 | 0.001 | 0.014 | 0.161 | 0.049 |
|  | 4 | - | - | - | - | - | - | - | - | 0.035 | - |
| Mirror dory | 1 | 0.534 | 0.476 | 0.852 | 0.333 | - | 0.385 | 0.353 | 0.398 | 0.641 | 0.333 |
|  | 2 | - | - | 0.680 | 0.274 | - | 0.302 | - | - | 0.484 | 0.527 |
|  | 3 | - | 0.625 | 0.115 | 0.232 | 0.101 | 0.356 | 0.240 | 0.114 | 0.180 | 0.072 |
|  | 4 | - | - | - | - | - | - | - | - | - | - |
| King dory | 1 | - | - | - | 0.162 | 0.105 | 0.122 | 0.069 | 0.029 | 0.038 | 0.048 |
|  | 2 | - | - | - | - | - | - | - | - | - | - |
|  | 3 | - | - | 0.046 | 0.090 | 0.102 | 0.245 | 0.098 | 0.060 | 0.047 | 0.017 |
|  | 4 | - | - | - | - | - | 0.032 | 0.229 | - | - | - |
| Western gemfish | 3 | - | 0.067 | 0.058 | 0.084 | 0.204 | 0.044 | 0.009 | 0.058 | 0.035 | 0.059 |
|  | 4 | - | - | - | - | - | - | - | - | - | - |


| O | (b) Coefficients of variation |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Fleet | 1993 | 1994 | 1995 | 1996 | 1997 | 1998 | 1999 | 2000 | 2001 | 2002 |
|  | Blue grenadier | 1 | 0.710 | 1.082 | 0.602 | 0.243 | 0.105 | 0.375 | 0.863 | 0.359 | 0.238 | 0.361 |
|  |  | 2 | - | - | - | - | 0.235 | - | 0.480 | 0.620 | 0.606 | - |
|  |  | 3 | - | 0.495 | 0.353 | 0.192 | 0.075 | 0.139 | 0.343 | 0.320 | 0.404 | 0.523 |
|  |  | 4 | - | - | - | - | - | 0.494 | 0.522 | - | - | 1.372 |
|  | Eastern gemfish | 1 | 0.668 | 0.675 | 0.512 | 0.734 | 0.314 | 0.559 | 0.388 | 0.427 | 0.411 | 0.334 |
|  |  | 2 | 0.072 | - | 0.675 | 0.508 | 0.552 | 0.404 | 0.347 | 0.365 | 0.406 | 0.408 |
|  | Pink ling | 1 | 0.610 | 0.413 | 0.336 | 0.245 | 0.248 | 0.241 | 0.284 | 0.405 | 0.395 | 0.464 |
|  |  | 2 | 1.097 | 0.645 | 0.395 | 0.478 | 0.244 | 0.563 | 0.662 | - | 0.794 | 0.503 |
|  |  | 3 | - | 1.045 | 0.633 | 0.683 | 0.598 | 0.373 | 0.616 | 0.641 | 0.402 | 0.569 |
|  |  | 4 | - | - | - | - | - | 0.882 | 1.014 | - | - | - |
|  | Spotted warehou | 1 | 0.685 | 0.524 | 0.450 | 0.702 | 0.467 | 0.473 | 0.472 | 0.540 | 0.486 | 0.468 |
|  |  | 2 | 0.463 | 0.460 | 0.304 | 0.533 | 0.506 | 0.987 | 1.085 | 0.851 | 0.416 | 0.891 |
|  |  | 3 | - | 0.687 | 0.718 | 0.220 | 0.524 | 0.197 | 0.540 | 0.543 | 0.358 | 0.350 |
|  |  | 4 | - | - | - | - | - | 0.531 | 0.823 | - | 0.416 | - |
|  | Ocean perch | 1 | 0.238 | 0.104 | 0.070 | 0.092 | 0.221 | 0.199 | 0.160 | 0.162 | 0.076 | 0.245 |
|  |  | 2 | 0.436 | 0.299 | 0.169 | 0.224 | 0.314 | 0.419 | 0.382 | 0.142 | 0.095 | 0.245 |
|  |  | 3 | - | 0.415 | 0.544 | 0.711 | 0.632 | 0.517 | 0.996 | 0.587 | 0.398 | 0.445 |
|  |  | 4 | - | - | - | - | - | - | - | - | 0.519 | - |
|  | Mirror dory | 1 | 0.109 | 0.111 | 0.072 | 0.198 | - | 0.159 | 0.127 | 0.162 | 0.090 | 0.142 |
|  |  | 2 | - | - | 0.154 | 0.214 | - | 0.272 | - | - | 0.220 | 0.150 |
|  |  | 3 | - | 0.117 | 0.139 | 0.133 | 0.157 | 0.107 | 0.133 | 0.233 | 0.166 | 0.163 |
|  |  | 4 | - | - | - | - | - | - | - | - | - | - |
|  | King dory | 1 | - | - | - | 0.303 | 0.220 | 0.454 | 0.166 | 0.125 | 0.466 | 0.188 |
|  |  | 2 | - | - | - | - | - | - | - | - | - | - |
|  |  | 3 | - | - | 0.211 | 0.268 | 0.191 | 0.166 | 0.231 | 0.627 | 0.187 | 0.202 |
|  |  | 4 | - | - | - | - | - | 0.785 | 0.271 | - | - | - |
|  | Western gemfish | 3 | - | 0.323 | 0.523 | 0.402 | 0.550 | 0.453 | 0.342 | 0.444 | 0.574 | 0.228 |
|  |  | 4 | - | - | - | - | - | - | - | - | - | - |

Table 23. Standardized catch-rate indices by stock and fleet.

|  | Eastern gemfish |  | Blue grenadier |  | Pink ling |  |  |  | Spotted warehou |  |  |  | Western gemfish Fleet 3 | $\begin{gathered} \hline \begin{array}{c} \text { Mirror } \\ \text { dory } \end{array} \\ \hline \text { Fleet } 2 \end{gathered}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Fleet 2 | Fleet 2 | Fleet 3 | Fleet 4 | Fleet 1 | Fleet 2 | Fleet 3 | Fleet 4 | Fleet 1 | Fleet 2 | Fleet 3 | Fleet 4 |  |  |
| 1973 | 2811 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 1974 | 3082 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 1975 | 2533 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 1976 | 3440 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 1977 | 3237 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 1978 | 3562 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 1979 | 2780 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 1980 | 3127 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 1981 | 2717 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 1982 | 2100 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 1983 | 1163 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 1984 | 1259 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 1985 | 1537 |  | 43.36 | 282.71 | 3.92 | -1.00 | 23.26 |  |  | 16.11 |  | 177.82 | 2.507 |  |
| 1986 | 1414 | 7.244 | 34.16 | 173.12 | 4.05 | 3.90 | 17.54 |  | 2.99 | 11.75 | 35.58 | 102.59 | 2.642 | 0.939 |
| 1987 | 1766 | 8.617 | 40.44 | 182.47 | 4.10 | 4.18 | 21.53 |  | 10.48 | 19.17 | 49.20 | 129.70 | 2.742 | 1.148 |
| 1988 |  | 10.415 | 48.72 | 383.16 | 4.35 | 5.10 | 18.82 | 6.60 | 11.23 | 17.31 | 42.77 | 153.87 | 2.695 | 1.114 |
| 1989 | 1294 | 5.841 | 46.63 | 105.49 | 3.92 | 4.35 | 21.21 |  | 4.23 | 19.48 | 25.06 | 122.76 | 2.415 | 1.604 |
| 1990 | 1165 | 4.228 | 39.25 | 168.78 | 4.31 | 7.10 | 18.99 |  | 11.52 | 14.95 | 49.90 | 85.33 | 1.513 | 0.946 |
| 1991 | 930 | 3.406 | 43.43 | 407.25 | 3.88 | 6.19 | 17.97 |  | 11.00 | 9.33 | 32.12 | 87.32 | 1.562 | 1.278 |
| 1992 |  | 2.764 | 44.82 | 256.51 | 3.90 | 5.92 | 13.74 |  | 11.97 | 13.28 | 12.53 | 71.15 | 1.128 | 1.026 |
| 1993 |  | 1.773 | 29.35 | 355.49 | 4.37 | 5.61 | 18.90 |  | 11.64 | 12.09 | 21.23 | 89.61 | 1.018 | 0.991 |
| 1994 |  | 1.175 | 27.16 | 280.64 | 4.34 | 6.36 | 22.62 | 4.07 | 11.48 | 14.43 | 31.58 | 83.03 | 1.149 | 0.764 |
| 1995 |  | 0.769 | 18.67 | 135.83 | 4.68 | 9.32 | 22.99 | 3.05 | 9.58 | 12.30 | 30.37 | 64.51 | 1.022 | 0.649 |
| 1996 | 1371 | 0.950 | 16.07 | 198.12 | 4.09 | 6.30 | 23.65 | 4.63 | 7.53 | 11.33 | 15.47 | 80.29 | 1.073 | 0.740 |
| 1997 | 643 | 1.389 | 13.26 | 166.19 | 3.80 | 7.22 | 25.47 | 4.89 | 5.41 | 10.65 | 17.50 | 95.11 | 1.000 | 0.821 |
| 1998 | 926 | 1.000 | 26.19 | 181.50 | 3.86 | 7.65 | 26.14 | 3.74 | 15.45 | 8.54 | 17.31 | 113.52 | 1.059 | 1.000 |
| 1999 |  | 0.803 | 26.93 | 151.85 | 3.58 | 8.45 | 20.47 | 2.51 | 10.40 | 6.18 | 19.81 | 99.99 | 1.029 | 0.936 |
| 2000 |  | 0.545 | 23.69 | 178.37 | 2.77 | 6.48 | 19.03 | 3.39 | 26.13 | 6.67 | 17.53 | 84.06 | 1.137 | 0.707 |
| 2001 |  | 0.623 | 12.50 | 294.79 | 2.18 | 4.74 | 17.82 | 2.85 | 18.19 | 6.93 | 18.55 | 65.30 | 0.847 | 0.902 |
| 2002 |  | 0.358 | 14.75 | 210.42 | 1.84 | 4.02 | 15.87 | 2.39 | 16.73 | 9.01 | 15.12 | 67.17 |  | 0.643 |

Table 24. Number of catch-effort records by fleet and the number used when standardizing the catch and effort data.

|  | Total records |  |  |  | Blue grenadier |  | Pink ling |  |  |  | Spotted warehou |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Fleet 1 | Fleet 2 | Fleet 3 | Fleet 4 | Fleet 3 | Fleet 4 | Fleet 1 | Fleet 2 | Fleet 3 | Fleet 4 | Fleet 1 | Fleet 2 | Fleet 3 | Fleet 4 |
| 1985 | 15567 | 430 | 1334 | 94 | 452 | 40 | 442 | 0 | 174 | 0 | 27 | 0 | 323 | 0 |
| 1986 | 46337 | 15404 | 4265 | 269 | 1345 | 74 | 1427 | 442 | 656 | 0 | 124 | 205 | 881 | 16 |
| 1987 | 41571 | 14352 | 5519 | 441 | 1369 | 152 | 1380 | 556 | 578 | 0 | 93 | 154 | 434 | 41 |
| 1988 | 45731 | 15907 | 5756 | 191 | 1401 | 89 | 1185 | 564 | 611 | 11 | 248 | 330 | 433 | 33 |
| 1989 | 45930 | 16118 | 5712 | 210 | 1771 | 24 | 1438 | 615 | 897 | 0 | 459 | 125 | 528 | 4 |
| 1990 | 39990 | 13321 | 7100 | 823 | 1200 | 126 | 1329 | 391 | 931 | 0 | 335 | 293 | 456 | 30 |
| 1991 | 33634 | 11270 | 6223 | 326 | 2067 | 85 | 1718 | 474 | 1506 | 0 | 616 | 308 | 569 | 24 |
| 1992 | 24844 | 7860 | 5161 | 322 | 2009 | 185 | 1908 | 634 | 1533 | 0 | 681 | 227 | 506 | 32 |
| 1993 | 25412 | 9353 | 6185 | 412 | 2377 | 129 | 2794 | 919 | 2098 | 0 | 1144 | 536 | 1321 | 32 |
| 1994 | 28135 | 8967 | 6011 | 622 | 2075 | 298 | 3505 | 1131 | 1762 | 31 | 1716 | 610 | 1330 | 111 |
| 1995 | 25238 | 8101 | 8061 | 787 | 3144 | 406 | 3808 | 1326 | 2931 | 145 | 1664 | 522 | 1394 | 250 |
| 1996 | 28507 | 10003 | 8113 | 945 | 2966 | 334 | 4372 | 1110 | 2843 | 234 | 2533 | 743 | 1325 | 185 |
| 1997 | 30133 | 10123 | 8070 | 769 | 3131 | 364 | 4709 | 1399 | 3327 | 294 | 2139 | 719 | 1830 | 122 |
| 1998 | 30367 | 9679 | 7624 | 821 | 3530 | 538 | 4254 | 1274 | 3441 | 288 | 1460 | 527 | 1739 | 184 |
| 1999 | 30412 | 10450 | 8028 | 1250 | 4167 | 824 | 4460 | 1644 | 3407 | 362 | 1210 | 701 | 2367 | 420 |
| 2000 | 27181 | 9380 | 8999 | 1165 | 4617 | 918 | 4581 | 1573 | 3926 | 487 | 1227 | 997 | 2792 | 600 |
| 2001 | 23998 | 8498 | 9673 | 1374 | 4198 | 1030 | 4091 | 1306 | 4061 | 517 | 1119 | 696 | 3154 | 591 |
| 2002 | 24227 | 8808 | 9187 | 1296 | 4219 | 942 | 3456 | 1264 | 3985 | 434 | 1480 | 740 | 3532 | 664 |

Table 24. (continued) Number of catch-effort records by fleet and the number used when standardizing the catch and effort data.

|  | Total records |  |  |  |  | Eastern gemfish | Western gemfish |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Fleet 1 | Fleet 2 | Fleet 3 | Fleet 4 | Fleet 2 | Fleet 3 | Fleet 2 |
| 1985 | 15567 | 430 | 1334 | 94 | 0 | 481 | 0 |
| 1986 | 46337 | 15404 | 4265 | 269 | 1344 | 1316 | 244 |
| 1987 | 41571 | 14352 | 5519 | 441 | 1610 | 915 | 271 |
| 1988 | 45731 | 15907 | 5756 | 191 | 1084 | 995 | 331 |
| 1989 | 45930 | 16118 | 5712 | 210 | 901 | 857 | 322 |
| 1990 | 39990 | 13321 | 7100 | 823 | 503 | 767 | 196 |
| 1991 | 33634 | 11270 | 6223 | 326 | 286 | 1282 | 100 |
| 1992 | 24844 | 7860 | 5161 | 322 | 321 | 701 | 128 |
| 1993 | 25412 | 9353 | 6185 | 412 | 768 | 696 | 216 |
| 1994 | 28135 | 8967 | 6011 | 622 | 680 | 768 | 226 |
| 1995 | 25238 | 8101 | 8061 | 787 | 469 | 902 | 189 |
| 1996 | 28507 | 10003 | 8113 | 945 | 731 | 1168 | 230 |
| 1997 | 30133 | 10123 | 8070 | 769 | 756 | 1086 | 282 |
| 1998 | 30367 | 9679 | 7624 | 821 | 549 | 1427 | 184 |
| 1999 | 30412 | 10450 | 8028 | 1250 | 627 | 1353 | 296 |
| 2000 | 27181 | 9380 | 8999 | 1165 | 645 | 1283 | 255 |
| 2001 | 23998 | 8498 | 9673 | 1374 | 382 | 1135 | 184 |
| 2002 | 24227 | 8808 | 9187 | 1296 | 461 |  | 247 |

Table 25. Port length-frequency sample sizes. The underlined values indicated were not extracted from raw length-frequency information.

The sample sizes for blue grenadier (1984-89) are set to nominal values of 500.
(a) Sex-aggregated

|  |  | 1975 | 1976 | 1977 | 1978 | 1979 | 1980 | 1981 | 1982 | 1983 | 1984 | 1985 | 1986 | 1987 | 1988 | 1989 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| B. grenadier | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
|  | 2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
|  | 3 | 0 | 0 | 0 | 0 | 164 | 40 | 1425 | 478 | 0 | 500 | 500 | 0 | 500 | 500 | 500 |
|  | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 500 | 500 | 0 | 0 | 500 | 0 |
| E. gemfish | 1 | $\underline{2550}$ | 417 | 2365 | 3069 | 7360 | 5542 | 3160 | $\underline{1440}$ | $\underline{4063}$ | $\underline{2250}$ | 761 | 1691 | 441 | 1238 | 1349 |
|  | 2 | $\underline{1412}$ | $\underline{2178}$ | $\underline{7490}$ | $\underline{5407}$ | $\underline{5311}$ | $\underline{10075}$ | $\underline{16865}$ | $\underline{9843}$ | $\underline{8900}$ | 6512 | 7709 | 6718 | $\underline{8046}$ | 8999 | $\underline{9359}$ |
| Pink ling | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | $\underline{1957}$ | $\underline{4723}$ | $\underline{1233}$ | $\underline{0}$ |
|  | 2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | $\underline{432}$ | 718 | 1137 | $\underline{59}$ |
|  | 3 | 0 | 0 | 0 | 0 | 114 | 86 | 602 | 120 | 0 | 0 | 0 | 0 | $\underline{4007}$ | $\underline{3681}$ | 0 |
|  | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
|  | 5 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| S. warehou | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
|  | 2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
|  | 3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | $\underline{1194}$ | 854 | 110 |
|  | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| M. dory | 1 | 37 | 348 | $\underline{2603}$ | 555 | 0 | 0 | 0 | 62 | 189 | 499 | 109 | 577 | $\underline{2415}$ | 0 | 0 |
|  | 2 | 629 | 669 | $\underline{1248}$ | $\underline{1808}$ | 0 | 0 | 0 | $\underline{3045}$ | $\underline{2519}$ | $\underline{1759}$ | $\underline{2052}$ | $\underline{2681}$ | $\underline{2915}$ | 0 | 0 |
|  | 3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
|  | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| O. perch | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
|  | 2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
|  | 3 | 0 | 0 | 0 | 0 | 32 | 168 | 461 | 118 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
|  | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| K. dory | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
|  | 2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
|  | 3 | 0 | 0 | 0 | 0 | 30 | 0 | 47 | 40 | 0 | 0 | 0 | 0 | $\underline{2546}$ | $\underline{1659}$ | 697 |
|  | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| W. gemfish | 3 | 0 | 0 | 0 | 0 | 0 | 105 | 1399 | 121 | 0 | 0 | 0 | 0 | $\underline{21273}$ | 15200 | 0 |
|  | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |

Table 25 (continued)
(a) Sex-aggregated (continued)

|  | 1990 | 1991 | 1992 | 1993 | 1994 | 1995 | 1996 | 1997 | 1998 | 1999 | 2000 | 2001 | 2002 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| B. grenadier | 0 | 394 | 1307 | 1050 | 433 | 50 | 567 | 631 | 1388 | 4579 | 3112 | 4114 | 1858 |
|  | 0 | 74 | 117 | 0 | 29 | 0 | 0 | 477 | 190 | 1353 | 1068 | 398 | 841 |
|  | 0 | 459 | 2408 | 437 | 8162 | 6888 | 4830 | 10083 | 14656 | 7354 | 9433 | 7447 | 6717 |
|  | 0 | 0 | 774 | 0 | 1038 | 465 | 927 | 851 | 1648 | 10076 | 5629 | 7876 | 2476 |
| E. gemfish | 127 | 27 | 362 | 3483 | 1648 | 5281 | 1485 | 2244 | 2850 | 1672 | 1613 | 3107 | 1567 |
|  | 4938 | 4870 | 4598 | 3012 | 0 | 4433 | 0 | 101 | 347 | 1592 | 994 | 3007 | 1344 |
| Pink ling | 402 | 100 | 0 | 0 | 0 | 57 | 0 | 102 | 928 | 1120 | 334 | 338 | 1945 |
|  | 0 | 0 | 54 | 0 | 0 | 191 | 0 | 0 | 489 | 362 | 155 | 7 | 868 |
|  | 0 | 0 | 399 | 0 | 0 | 784 | 1180 | 2340 | 1311 | 925 | 1364 | 2293 | 1829 |
|  | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
|  | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 4021 | 10319 |
| S. warehou | 0 | 246 | 439 | 542 | 110 | 241 | 883 | 1114 | 4272 | 4464 | 3949 | 6907 | 9254 |
|  | 0 | 27 | 1209 | 545 | 105 | 259 | 131 | 648 | 2114 | 1883 | 3466 | 1051 | 3725 |
|  | 0 | 51 | 1630 | 1702 | 1761 | 4002 | 5715 | 8874 | 9500 | 7799 | 5201 | 6849 | 9064 |
|  | 0 | 0 | 139 | 40 | 41 | 649 | 308 | 0 | 204 | 50 | 106 | 129 | 0 |
| M. dory | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 347 | 593 | 250 | 1936 | 710 |
|  | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 44 | 79 | 124 | 289 | 459 |
|  | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 164 | 0 | 0 | 0 | 36 | 103 |
|  | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| O. perch | 0 | 0 | 0 | 214 | 0 | 0 | 0 | 0 | 1497 | 4007 | 3186 | 6107 | 2775 |
|  | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 628 | 1776 | 855 | 286 | 834 |
|  | 0 | 0 | 0 | 159 | 0 | 0 | 133 | 0 | 0 | 154 | 0 | 0 | 0 |
|  | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| K. dory | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 297 | 218 | 0 | 27 |
|  | 0 | 25 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 91 | 0 | 0 | 0 |
|  | 0 | 112 | 0 | 95 | 0 | 0 | 87 | 334 | 0 | 64 | 124 | 820 | 953 |
|  | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| W. gemfish | 0 | 126 | 463 | 1605 | 8153 | 2495 | 4223 | 4604 | 5551 | 3312 | 5803 | 5263 | 3939 |
|  | 0 | 0 | 0 | 0 | 76 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |

Table 25 (continued)
(b) Males

|  |  | 1975 | 1976 | 1977 | 1978 | 1979 | 1980 | 1981 | 1982 | 1983 | 1984 | 1985 | 1986 | 1987 | 1988 | 1989 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| B. grenadier | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| E. gemfish | 2 | 0 | 0 | 0 | 0 | 0 | $\underline{4672}$ | 0 | $\underline{4702}$ | $\underline{4107}$ | $\underline{3120}$ | 3841 | $\underline{3065}$ | $\underline{4152}$ | $\underline{4932}$ | $\underline{3936}$ |
| Pink ling | 5 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |


|  | 1990 | 1991 | 1992 | 1993 | 1994 | 1995 | 1996 | 1997 | 1998 | 1999 | 2000 | 2001 | 2002 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| B. grenadier | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | $\underline{5929}$ | $\underline{2957}$ | $\underline{4255}$ | 1220 |
| E. gemfish | $\underline{2115}$ | $\underline{1893}$ | $\underline{1450}$ | 1134 | 1988 | $\underline{2290}$ | $\underline{6014}$ | $\underline{3105}$ | $\underline{5264}$ | 1186 | 936 | 0 | 0 |
| Pink ling | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 720 |

## (c) Females

|  |  | 1975 | 1976 | 1977 | 1978 | 1979 | 1980 | 1981 | 1982 | 1983 | 1984 | 1985 | 1986 | 1987 | 1988 | 1989 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| B. grenadier | 4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| E. gemfish | 2 | 0 | 0 | 0 | 0 | 0 | 5403 | 0 | $\underline{5143}$ | $\underline{4794}$ | $\underline{3392}$ | 3869 | $\underline{3655}$ | 3894 | 4067 | 5425 |
| Pink ling | 5 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |


|  | 1990 | 1991 | 1992 | 1993 | 1994 | 1995 | 1996 | 1997 | 1998 | 1999 | 2000 | 2001 | 2002 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| B. grenadier | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 4147 | $\underline{2672}$ | 3619 | 1255 |
| E. gemfish | $\underline{2823}$ | $\underline{2977}$ | $\underline{3150}$ | $\underline{2297}$ | $\underline{1119}$ | $\underline{2146}$ | $\underline{5488}$ | $\underline{4457}$ | 4609 | $\underline{1033}$ | $\underline{666}$ | 0 | 0 |
| Pink ling | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 694 |

Table 26. Specifications for the pre-specified parameters of the population dynamics model for each of the eight stocks.

| Quantity | Blue grenadier ${ }^{1}$ |  | Eastern gemfish |  |
| :---: | :---: | :---: | :---: | :---: |
|  | Females | Males | Females | Males |
| Natural mortality, M (yr ${ }^{-1}$ ) | 0.2 | 0.25 | $0.4{ }^{2}$ | $0.6{ }^{2}$ |
| Growth parameters |  |  |  |  |
| $\ell_{\infty}(\mathrm{cm})$ | 94.0 | 94.0 | $109.4{ }^{3}$ | $97.5^{3}$ |
| $\kappa\left(\mathrm{yr}^{-1}\right)$ | 0.2219 | 0.2219 | $0.18^{3}$ | $0.212^{3}$ |
| $t_{0}$ (yr) | -1.09 | -1.09 | $-0.63{ }^{3}$ | $-0.54{ }^{3}$ |
| $\tilde{\sigma}_{0}$ | 0.173 | 0.173 | 0.116 | 0.141 |
| $\tilde{\sigma}_{50}$ | 0.134 | 0.123 | 0.087 | 0.097 |
| Length-weight - $a$ | 0.00375 | 0.00375 | $0.00143^{4}$ | $0.00143^{4}$ |
| Length-weight - b | 3.013 | 3.013 | 3.394 | 3.394 |
| Age-at-maturity, $a_{m}$ | 5 |  | A |  |
| Fraction spawning, $\mu$ | 0.77 |  | 1 |  |
| Maximum age, $x$ |  |  |  |  |
| Extent of recruitment variation, $\sigma_{R}$ |  |  |  |  |
| Stock-recruitment steepness, $h^{\text {s }}$ |  |  |  |  |
| Discard-related parameters |  |  |  |  |
| Length-at-50\%-retention, $\phi_{50}$ |  |  |  |  |
| Length-at-50\%-retention, $\phi_{95}$ |  |  |  |  |

1 - Punt et al. (2001a)
2 - Smith and Punt (1998)
3 - Rowling and Reid (1992)
4 - K. Rowling (NSW FRI, pers. commn)
A - Assumed to be the same as selectivity to the winter fishery (Smith and Punt, 1998)

Table 26 (continued)

| Quantity | Pink Ling ${ }^{1}$ |  | Spotted warehou ${ }^{2}$ |  |
| :---: | :---: | :---: | :---: | :---: |
|  | Females | Males | Females | Males |
| Natural mortality, $M$ ( $\mathrm{yr}^{-1}$ ) | 0.2 | 0.2 | 0.25 | 0.25 |
| Growth parameters |  |  |  |  |
| $\ell_{\infty}(\mathrm{cm})$ | 109.96 | 109.96 | 50.69 | 50.69 |
| $\kappa\left(\mathrm{yr}^{-1}\right)$ | 0.135 | 0.135 | 0.372 | 0.372 |
| $t_{0}$ (yr) | -2.72 | -2.72 | -0.451 | -0.451 |
| $\tilde{\sigma}_{0}$ | 0.188 | 0.199 | 0.175 | 0.170 |
| $\tilde{\sigma}_{50}$ | 0.145 | 0.137 | 0.045 | 0.030 |
| Length-weight - $a$ | 0.0293 | 0.0293 | 0.0065 | 0.0065 |
| Length-weight - b | 3.139 | 3.139 | 3.27 | 3.27 |
| Age-at-maturity, $a_{m}$ | 5 (67cm) |  | 3 (37cm) |  |
| Fraction spawning, $\mu$ | 1 |  | 1 |  |
| Maximum age, $x$ |  |  |  |  |
| Extent of recruitment variation, $\sigma_{R}$ |  |  |  |  |
| Stock-recruitment steepness, $h^{\text {s }}$ |  |  |  |  |
| Discard-related parameters |  |  |  |  |
| Length-at-50\%-retention, $\phi_{50}$ | 40.58 cm |  | 30.48 cm |  |
| Length-at-50\%-retention, $\phi_{95}$ | 31.83 cm |  | 22.63 cm |  |
| 1 - Klaer (2003) |  |  |  |  |
| 2 - Thomson (2002) |  |  |  |  |

Table 26 (continued)

| Quantity | Mirror dory |  | Ocean perch |  |
| :---: | :---: | :---: | :---: | :---: |
|  | Females | Males | Females | Males |
| Natural mortality, $M\left(\mathrm{yr}^{-1}\right)$ | $0.38{ }^{1}$ | 0.38 v | $0.12{ }^{2}$ | $0.1^{2}$ |
| Growth parameters |  |  |  |  |
| $\ell_{\infty}(\mathrm{cm})$ | 60.91 | 60.91 | $42.87^{2}$ | $42.87^{2}$ |
| $\kappa\left(\mathrm{yr}^{-1}\right)$ | $0.2^{1}$ | $0.2{ }^{1}$ | $0.07{ }^{2}$ | $0.07^{2}$ |
| $t_{0}$ (yr) | $0.18{ }^{1}$ | $0.18{ }^{1}$ | $-5.96{ }^{2}$ | $-5.96{ }^{2}$ |
| $\tilde{\sigma}_{0}$ | 0.099 | 0.092 | 0.077 | 0.087 |
| $\tilde{\sigma}_{50}$ | 0.099 | 0.092 | 0.077 | 0.087 |
| Length-weight - a | $0.0164^{4}$ | $0.0164^{4}$ | $0.0181^{2}$ | $0.0181^{2}$ |
| Length-weight - $b$ | $3^{4}$ | $3^{4}$ | $2.997^{2}$ | $2.997{ }^{2}$ |
| Age-at-maturity, $a_{m}$ | 2 |  | $31 \mathrm{~cm}^{3}(12.4)$ |  |
| Fraction spawning, $\mu$ | 1 |  | 1 |  |
| Maximum age, $x$ |  |  |  |  |
| Extent of recruitment variation, $\sigma_{R}$ |  |  |  |  |
| Stock-recruitment steepness, $h^{\text {s }}$ |  |  |  |  |
| Discard-related parameters |  |  |  |  |
| Length-at-50\%-retention, $\phi_{50}$ |  |  |  |  |
| Length-at-50\%-retention, $\phi_{95}$ |  |  |  |  |
| 1 - Smith and Stewart (1994) |  |  |  |  |
| 2 - Smith and Wayte (2002) |  |  |  |  |
| 3 - Lyle and Ford (1993) |  |  |  |  |
| 4 - Robin Thomson (CSIRO, pers. commn) |  |  |  |  |
| 5 - Default values |  |  |  |  |

Table 26 (continued)

| Quantity | King dory |  | Western gemfish |  |
| :---: | :---: | :---: | :---: | :---: |
|  | Females | Males | Females | Males |
| Natural mortality, $M$ ( $\mathrm{yr}^{-1}$ ) | $0.1^{1}$ | $0.1^{1}$ | $0.25{ }^{3}$ | $0.25^{3}$ |
| Growth parameters |  |  |  |  |
| $\ell_{\infty}(\mathrm{cm})$ | $51.36{ }^{1}$ | $51.36{ }^{1}$ | $98.12^{2}$ | $98.12^{2}$ |
| $\kappa\left(\mathrm{yr}^{-1}\right)$ | $0.1^{1}$ | $0.1^{1}$ | $0.19^{2}$ | $0.19{ }^{2}$ |
| $t_{0}$ (yr) | $-0.83{ }^{1}$ | $-0.83^{1}$ | $-1.0^{2}$ | $-1.0^{2}$ |
| $\tilde{\sigma}_{0}$ | 0.059 | 0.077 | 0.128 | 0.142 |
| $\tilde{\sigma}_{50}$ | 0.059 | 0.077 | 0.097 | 0.106 |
| Length-weight - $a$ | $0.0771^{2}$ | $0.0771^{2}$ | $0.072^{2}$ | 0.072 ${ }^{2}$ |
| Length-weight - b | $3.148^{2}$ | $3.148^{2}$ | $3.00^{2}$ | $3.00^{2}$ |
| Age-at-maturity, $a_{m}$ | $\begin{gathered} 12.5 \\ (35-40 \mathrm{~cm}) \end{gathered}$ |  | $60-70 \mathrm{~cm}$ |  |
| Fraction spawning, $\mu$ | 1 |  | 1 |  |
| Maximum age, $x$ |  |  |  |  |
| Extent of recruitment variation, $\sigma_{R}$ |  |  |  |  |
| Stock-recruitment steepness, $h^{\text {s }}$ |  |  |  |  |
| Discard-related parameters |  |  |  |  |
| Length-at-50\%-retention, $\phi_{50}$ |  |  |  |  |
| Length-at-50\%-retention, $\phi_{95}$ |  |  |  |  |

1-Smith and Stewart (1994)
2 - D.C. Smith (MSFRI), pers. commn.
3 - Default values

Table 27. Number of animals aged by stock.

|  | Sex | 1980 | 1981 | 1982 | 1983 | 1984 | 1985 | 1986 | 1987 | 1988 | 1989 | 1990 | 1991 | 1992 | 1993 | 1994 | 1995 | 1996 | 1997 | 1998 | 1999 | 2000 | 2001 | 2002 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| B. grenadier | M +F | 0 | 0 | 0 | 0 | 767 | 603 | 174 | 514 | 531 | 502 | 0 | 193 | 1187 | 1362 | 1610 | 1344 | 1832 | 2338 | 2089 | 1574 | 2349 | 1953 | 1687 |
|  | M | 0 | 0 | 0 | 0 | 389 | 349 | 96 | 133 | 226 | 211 | 0 | 135 | 672 | 826 | 836 | 559 | 1009 | 834 | 724 | 900 | 1265 | 848 | 885 |
|  | F | 0 | 0 | 0 | 0 | 368 | 241 | 78 | 164 | 229 | 262 | 0 | 58 | 485 | 452 | 647 | 751 | 642 | 663 | 971 | 674 | 1073 | 990 | 802 |
| E. gemfish | M +F | 1278 | 0 | 1752 | 0 | 870 | 0 | 1297 | 0 | 562 | 0 | 554 | 1144 | 65 | 526 | 498 | 0 | 0 | 595 | 521 | 202 | 0 | 0 | 0 |
|  | M | 507 | 0 | 800 | 0 | 344 | 0 | 551 | 0 | 257 | 0 | 314 | 422 | 55 | 139 | 192 | 0 | 0 | 244 | 265 | 104 | 0 | 0 | 0 |
|  | F | 771 | 0 | 952 | 0 | 526 | 0 | 746 | 0 | 305 | 0 | 240 | 722 | 6 | 377 | 256 | 0 | 0 | 349 | 256 | 98 | 0 | 0 | 0 |
| Pink ling | M +F | 0 | 0 | 0 | 0 | 491 | 0 | 0 | 567 | 327 | 190 | 0 | 0 | 0 | 65 | 237 | 1110 | 1029 | 1163 | 879 | 1366 | 565 | 876 | 732 |
|  | M | 0 | 0 | 0 | 0 | 192 | 0 | 0 | 208 | 164 | 90 | 0 | 0 | 0 | 21 | 45 | 525 | 405 | 556 | 316 | 668 | 330 | 393 | 245 |
|  | F | 0 | 0 | 0 | 0 | 153 | 0 | 0 | 279 | 148 | 95 | 0 | 0 | 0 | 40 | 96 | 498 | 489 | 563 | 432 | 602 | 234 | 466 | 463 |
| S. warehou | M +F | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 456 | 140 | 0 | 0 | 0 | 0 | 370 | 359 | 451 | 515 | 566 | 585 | 903 | 470 | 997 | 627 |
|  | M | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 221 | 59 | 0 | 0 | 0 | 0 | 159 | 132 | 130 | 197 | 295 | 281 | 402 | 210 | 482 | 251 |
|  | F | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 235 | 76 | 0 | 0 | 0 | 0 | 184 | 158 | 292 | 270 | 267 | 271 | 452 | 258 | 510 | 340 |
| M. dory | M +F | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 399 | 77 | 0 | 0 |
|  | M | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 73 | 6 | 0 | 0 |
|  | F | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 325 | 71 | 0 | 0 |
| O. perch | M +F | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 507 | 111 | 121 | 0 |
|  | M | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 225 | 47 | 46 | 0 |
|  | F | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 231 | 64 | 74 | 0 |
| W. gemfish | M +F | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 10 | 133 | 133 | 498 | 312 | 0 | 0 | 0 | 0 | 464 | 494 |
|  | M | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 2 | 86 | 39 | 72 | 45 | 0 | 0 | 0 | 0 | 79 | 87 |
|  | F | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 8 | 17 | 33 | 346 | 122 | 0 | 0 | 0 | 0 | 383 | 399 |

Table 28. Number of otoliths for which multiple age-readings are available (by stock) and the estimates for the parameters of the model for age-reading error.

| Stock | Number of otoliths | CV-1 (age) | CV-2 (age) | CV-3 (age) |
| :--- | :---: | :---: | :---: | :---: |
| Blue grenadier | 3789 | $0.163(1)$ | $0.063(11)$ | $0.113(22)$ |
| Eastern gemfish | 372 | $0.242(1)$ | $0.109(6)$ | $0.341(13)$ |
| Pink ling | 1401 | $0.207(1)$ | $0.042(14)$ | $0.151(28)$ |
| Spotted warehou | 1114 | $0.161(1)$ | $0.098(9)$ | $0.145(18)$ |
| Mirror dory | 119 | $0.118(1)$ | $0.137(5)$ | $0.043(10)$ |
| Ocean perch | 217 | $0.121(1)$ | $0.021(20)$ | $0.220(40)$ |
| King dory | $\mathrm{N} / \mathrm{A}$ |  |  | $0.054(10)$ |
| Western gemfish | 401 | $0.175(1)$ | $0.116(5)$ | 0.0 |

Information for 'data-rich' species to inform assessments of 'data-poor' species

# Chapter 5: An Approach to Informing 'Data-Poor' Species Using Information for 'Data-Rich' Species 

## Introduction

Stock assessment methods are used to provide estimates of a variety of quantities for use in fisheries management. These quantities include historical and current biomass, the ratio of the current biomass to some historical, target or limit biomass, and the implications of future sequences of catches or levels of fishing effort. Stock assessment methods such as production models, yield-per-recruit analysis, and separable population analysis were designed for application to single stocks based on the data for the stock concerned and for that stock only. In recent years, however, two ways have arisen in which allowance can be made for the data for one stock to influence the results of stock assessments for other stocks.
The first of these ways relates to the use in assessments of prior probability distributions. Prior probability distributions can be used to summarize the information about the value of a particular parameter based on the results of assessments of other stocks/species. Prior probability distributions are used most naturally and straightforwardly when conducting stock assessments using Bayesian methods (e.g. McAllister et al., 1994; Punt and Hilborn, 1997; McAllister and Kirkwood, 1998; Ianelli et al., 2000) although prior distributions can also be incorporated in non-Bayesian assessments by using them as penalty functions (Punt and Hilborn, 1996; Schweder and Ianelli, 1998) or as bounds (Punt and Butterworth, 2002). As noted in Chapter 3 Results and discussion - Impact of covariates, prior distributions based on the results of analyses for other stocks/species are already available for a variety of model parameters, including the steepness of the stockrecruitment relationship, the variability about the stock-recruitment relationship, and survey selectivity.
The second way in which the results from the assessments of one stock can influence those of another stock occurs when multiple stocks are assessed simultaneously. For example, Francis et al. (2002) assess the two stocks of blue grenadier off New Zealand within the same assessment framework because some of the indices of abundance (e.g. the surveys on the Chatham Rise) pertain to both stocks. Similarly, Punt et al. (2000) conducted assessments of the school shark, Galeorhinus galeus, off southern Australia using a population dynamics model that includes two stocks. Pribac et al. (in press) assess the gummy shark population, Mustelus antarcticus, off southern Australia using an assessment model in which some of the key population dynamics parameters (such as the Maximum Sustainable Yield rate, $M S Y R$ ) are the same for several stocks of gummy shark.

The approaches taken by Francis et al. (2002), Punt et al. (2000), and Pribac et al. (in press) all involve multiple stocks of the same species. This chapter extends this by assessing multiple stocks of several species simultaneously. It is clearly not possible to share values of parameters (such as the age-at-maturity) among different species. However, it seems plausible that if multiple stocks/species are exploited by the same fleet (where fleet is defined as a group of vessels fishing in the essentially the same fishing grounds at the same time), the trend in fishing mortality for that fleet should be similar for all of the stocks/species. This assumption is the same as that which underlies multi-species yield-per-recruit analysis (e.g. Murawski, 1996; Pikitch, 1987). It is not only trends in fishing mortality by fleet which might be expected to be common across stocks/species. It also seems likely that the annual deviations in recruitment about the stock-recruitment relationship for different stocks/species would be correlated (positively or negatively) due to the impact of common environmental variables, and that selectivity as a function of length (before discarding) should be relatively similar across various fleets.

Although it seems plausible that, for example, selectivity-at-length for a given fleet should be similar across stocks/species, there are also good reasons related to the behaviour of fishers and fish, why, for example, selectivity-at-length for different stocks/species should not be identical. The approach outlined below therefore allows for stock-specific values for all of the model parameters, but adds penalties on how different the values for these parameters may be across stocks.
The difference between the conventional approach to conducting fisheries stock assessments and that developed in this project is highlighted in Figs 48a and 48b. Figure 48a shows how assessments for multiple stocks are conducted conventionally - the objective function minimized to obtain the point estimates for the model parameters is the sum of stock-specific objective functions. As a result, the same answers would be obtained irrespective of whether the assessments are conducted simultaneously or separately. Figure 48b shows how the assessments of this chapter are conducted. The objective function minimized contains contributions for the fit to the data for each stock, but there are also contributions to the objective function related to the differences in selectivity-at-length among fleets, the deviations about the stock-recruitment relationship among stocks, and the pattern of fishing mortality over time among fleets. As a result, this approach to stock assessment has to be applied to all stocks simultaneously.

This chapter first outlines a multi-stock, multi-fleet population dynamics model and the objective function that is minimized to estimate the values for the parameters of this model. This objective function consists of contributions from the data available for assessment purposes and the penalties on inter-stock/species deviations in fishing mortality, selectivity and the recruitment deviations. This model is then fitted to the data for eight stocks (based on seven species) in Australia's SESS ${ }^{3}$. Results are shown based on minimizing the objective function and by applying the Markov Chain Monte Carlo algorithm to generate a set of equally likely parameter values from the posterior distribution.

## The population dynamics model

## Basic dynamics

The dynamics of the each of the stocks included in the assessment are assumed to be governed by the standard age-structured equations, i.e.:


[^4]$x$
is the maximum age-class (treated as a plus-group) ${ }^{4}$.
a) Conventional approach


The objective function is the sum of the contributions by data for each stock
b) Multi-stock approach of this report


The objective function is the sum of the contributions by data for each stock plus inter-stock differences in fishing pattern, selectivity and recruitment

Figure 48. Comparison of how the objective function is developed for the conventional approach to fisheries stock assessment and that of the approach of this paper.

[^5]
## Stock and recruitment

The number of zero-year-olds added to the population each year is assumed to be governed by a stochastic version of the Beverton-Holt stock-recruitment relationship:

$$
\begin{equation*}
R_{y}^{s}=\frac{4 h^{s} R_{0}^{s} S_{y}^{s}}{\left(1-h^{s}\right) S_{0}^{s}+\left(5 h^{s}-1\right) S_{y}^{s}} e^{\varepsilon_{y}^{s}-\left(\sigma_{R}^{s}\right)^{2} / 2} \tag{5.2}
\end{equation*}
$$

where $\quad R_{0}^{s}$ is the expected number of zero-year-olds (of both sexes) in stock $s$ in the absence of exploitation,
$S_{y}^{s} \quad$ is the spawning biomass of stock $s$ at the start of year $y$ :

$$
\begin{equation*}
S_{y}^{s}=\mu^{s} \sum_{a=1}^{x} f_{a}^{s} N_{y, a}^{\mathrm{f}, s} \tag{5.3}
\end{equation*}
$$

$f_{a}^{s} \quad$ is the fecundity of an animal of stock $s$ and age $a$,
$\mu^{s} \quad$ is the proportion of mature females of stock $s$ that spawn each year,
$h^{s} \quad$ is the steepness of the stock-recruitment relationship (the fraction of $R_{0}^{s}$ to be expected when the spawning biomass is reduced to $0.2 S_{0}^{s}$ ) for stock $s$,
$\varepsilon_{y}^{s} \quad$ is the deviation during year $y$ about the stock-recruitment relationship for stock $s$ (the recruitment residual for stock $s$ and year $y$ ), and
$\sigma_{R}^{s}$ is the extent of variability in recruitment about the stock-recruitment relationship.

## Catches

The total catch (in numbers) by fleet, year, stock, sex and age-class is given by:

$$
\begin{equation*}
C_{y, a}^{g, s, f}=F_{y}^{s, f} S_{a}^{g, s, f}\left(N_{y, a}^{g, s} e^{-M^{g, s / 2}}-\sum_{f^{\prime}<f} C_{y, a}^{g, s, f^{\prime}}\right) \tag{5.4}
\end{equation*}
$$

where $\quad S_{a}^{g, s, f}$ is the selectivity of fleet $f$ on animals of stock $s$, sex $g$ and age $a$ (assumed to be timeinvariant),
$F_{y}^{s, f} \quad$ is the exploitation rate on fully selected (i.e. $S_{a}^{g, s, f} \rightarrow 1$ ) animals of stock $s$ by fleet $f$ during year $y$, i.e.:

$$
\begin{equation*}
F_{y}^{s, f}=\tilde{C}_{y}^{s, f} / B_{y}^{e, s, f} \tag{5.5}
\end{equation*}
$$

[^6]$B_{y}^{e, s, f} \quad$ is the exploitable biomass of stock $s$ in year $y$ prior to removal of the catch by fleet $f$ :
\[

$$
\begin{equation*}
B_{y}^{e, s, f}=\sum_{g} \sum_{a=0}^{x} w_{a+1 / 2}^{g, s}\left(1-\phi_{a}^{g, s, f}\right) S_{a}^{g, s, f}\left(N_{y, a}^{g, s} e^{-M^{g, s / 2}}-\sum_{f^{\prime}<f} C_{y, a}^{g, s, f}\right) \tag{5.6}
\end{equation*}
$$

\]

$\tilde{C}_{y}^{s, f}$ is total catch (in weight) by fleet, year and sex,
$w_{a+1 / 2}^{g, s}$ is the weight of a fish of stock $s$, sex $g$, and age $a$ in the middle of the year, and
$\phi_{a}^{g, s, f} \quad$ is the fraction of the catch by fleet $f$ of animals of stock $s$, sex $g$ and age $a$ that is discarded.
The model (Equations 5.1 and 5.6) assumes that the catch is taken in a pulse in the middle of the year and that the catches by each fleet are taken sequentially (i.e. the fisheries are assumed to be gauntlet fisheries). These assumptions are made to avoid having to treat the fully-selected exploitation rate for each year, fleet and stock as estimable parameters.
The landed and discarded total catches (in numbers and weight respectively) are:

$$
\begin{array}{lr}
C_{y, a}^{\mathrm{L}, g, \mathrm{~s}, f}=\left(1-\phi_{a}^{g, \mathrm{~s}, f}\right) C_{y, a}^{g, s, f} ; & C_{y, a}^{D, g, s, f}=\phi_{a}^{g, s, f} C_{y, a}^{g, \mathrm{~s}, f} \\
\tilde{C}_{y}^{\mathrm{L}, \mathrm{~s}, f}=\sum_{g} \sum_{a=1}^{x} w_{a+1 / 2}^{g, s} C_{y, a}^{\mathrm{L}, g, s, f} ; & \tilde{C}_{y}^{\mathrm{D}, \mathrm{~s}, f}=\sum_{g} \sum_{a=1}^{x} w_{a+1 / 2}^{g, s} C_{y, a}^{\mathrm{D}, g, \mathrm{~s}, f} \tag{5.7b}
\end{array}
$$

## Selectivity and discarding

Two general selectivity ${ }^{5}$ patterns (logistic and normal) are available. These two patterns allow the analyst to specify that selectivity is either asymptotic (logistic) or domed-shaped (normal). The logistic selectivity model (Equation 5.8a) can be used to represent situations in which the probability of capturing an animal increases ( $L_{95}^{s, f}>L_{50}^{s, f}$ ) or decreases ( $L_{50}^{s, f}>L_{95}^{s, f}$ ) with length while two variants of the normal selectivity model (Equation 5.8 b ) are available depending on whether the selectivity pattern is assumed to be symmetric ( $\left.\Omega_{L}^{s, f}=\Omega_{R}^{s, f}\right)$ or $\operatorname{not}\left(\Omega_{L}^{s, f} \neq \Omega_{R}^{s, f}\right)$.

$$
\begin{equation*}
S_{a}^{g, s, f}=\left[1+\exp \left(-\ell \mathrm{n} 19 \frac{L_{a+1 / 2}^{g, S}-L_{50}^{s, f}}{L_{95}^{s, f}-L_{50}^{s, f}}\right)\right]^{-1} \tag{5.8a}
\end{equation*}
$$

where

$$
\begin{array}{ll}
L_{50}^{s, f} & \text { is the length-at- } 50 \% \text {-selectivity for fleet } f \text { on fish of stock } s, \\
L_{95}^{s, f} & \text { is the length-at- } 95 \% \text {-selectivity for fleet } f \text { on fish of stock } s \text {, and }
\end{array}
$$

[^7]```
La growth equation).
```

$$
S_{a}^{g, s, f}= \begin{cases}\exp \left[-\left(L_{a+1 / 2}^{s, f}-\bar{L}^{s, f}\right)^{2} / \Omega_{L}^{s, f}\right] & \text { if } L_{a+1 / 2}^{s, f} \leq \bar{L}^{s, f}  \tag{5.8b}\\ \exp \left[-\left(L_{a+1 / 2}^{s, f}-\bar{L}^{s, f}\right)^{2} / \Omega_{R}^{s, f}\right] & \text { otherwise }\end{cases}
$$

where $\bar{L}^{s, f} \quad$ is the length corresponding to maximum selectivity for fleet $f$ when fishing stock $s$,
$\Omega_{L}^{s, f} \quad$ is the parameter that determines how rapidly selectivity for fleet $f$ increases with length for stock $s$, and
$\Omega_{R}^{s, f} \quad$ is the parameter that determines how rapidly selectivity for fleet $f$ declines with length for stock $s$.

This formulation for selectivity assumes that the probability of capture is a function primarily of the length (rather than the age) of an animal. It also assumes that selectivity is only sex-specific to the extent that growth is sex-specific.
The probability of fleet $f$ discarding an animal of stock $s$, sex $g$ and age $a$ is assumed to be a function of length, i.e.:

$$
\begin{equation*}
\phi_{a}^{g, s, f}=\left[1+\exp \left(-\ln 19 \frac{L_{a}^{g_{a}, s}-\phi_{50}^{s, f}}{\left.\phi_{55}^{s, f}-\phi_{50}^{\phi_{5}}\right)}\right)\right]^{-1} \tag{5.9}
\end{equation*}
$$

where $\quad \phi_{50}^{s, f} \quad$ is the length at which $50 \%$ of animals of stock $s$ are discarded by fleet $f$,
$\phi_{95}^{s, f} \quad$ is the length at which $95 \%$ of animals of stock $s$ are discarded by fleet $f$.
The analyses of this chapter are based on the assumption that the probability of a fish being discarded as a function of length is independent of fleet. This assumption is based on the results in Figure 42.

## Initial conditions

The initial conditions correspond to a population at its deterministic unfished level with the corresponding age-structure, i.e.:

$$
N_{y_{1}, a}^{g, s}= \begin{cases}0.5 R_{0}^{s} e^{-a M^{g, s}} & \text { if } 0 \leq a<x  \tag{5.10}\\ 0.5 R_{0}^{s} e^{-x M^{g, s}} /\left(1-e^{-M^{g, s}}\right) & \text { if } a=x\end{cases}
$$

where $\quad y_{1} \quad$ is the first year for which catches are available.

## The objective function

The objective function includes contributions from the data available for assessment purposes (discard rates, fishery landed age-composition data, fishery landed size-composition data, survey estimates of female spawning biomass, and catch-rates) and priors (penalties) imposed on the recruitment residuals, selectivity, steepness, and fishing mortality.

## The likelihood function

The equations listed below assume that data for each data-type are available for every year, every stock and every fleet. This is not the case in reality and the equations are modified appropriately in the absence of data for specific years, stocks, and fleets.

## Discard rates

The contribution of the estimates of discard rate (in mass) by fleet, stock and year to the negative of the logarithm of the likelihood function is based on the assumption that the errors in measuring discard rate are log-normal, i.e.:

$$
\begin{equation*}
\ell \mathrm{n} L_{1}=\sum_{s} \sum_{f} \sum_{y}\left(\ell \mathrm{n} \sigma_{d}^{s, f}+\frac{1}{2\left(\sigma_{d}^{s, f}\right)^{2}}\left[\ell \mathrm{n} D_{y}^{s, f}-\ell \mathrm{n} D_{y}^{o b s, s, f}\right]^{2}\right) \tag{5.11}
\end{equation*}
$$

where $\boldsymbol{\nu}_{y} \quad$ is the model-estimate of the fraction of the catch of stock $s$ by fleet $f$ that was discarded during year $y$ :

$$
\begin{equation*}
D_{y}^{s, f}=\frac{\tilde{C}_{y}^{D, s, f}}{\tilde{C}_{y}^{L, s, f}+\tilde{C}_{y}^{D, s, f}} \tag{5.12}
\end{equation*}
$$

$D_{y}^{o b s, s, f}$ is the observed fraction of the catch (in mass) of stock $s$ by fleet $f$ that was discarded during year $y$ (Table 22a), and
$\sigma_{d}^{s, f} \quad$ is (approximately) the coefficient of variation of the discard rates for stock $s$ and fleet $f$ (Table 22b).

## Age- and size-composition data

The contribution of the age- and size-composition data (the proportion of the catch that is landed by age-/size-class by fleet, stock and year) to the negative of the logarithm of the likelihood function is based on the robust likelihood formulation of Fournier et al. (1990). The contribution of the age-/size-composition data for stock $s$, fleet $f$ and year $y$ to the negative of the logarithm of the likelihood function is therefore given by:

$$
\begin{equation*}
\ln L_{2}^{s, f, y}=-\sum_{i} \ell \mathrm{n}\left[\exp \left\{\frac{-\left(\rho_{i}-\rho_{i}^{o b s}\right)^{2}}{2\left[\left(1-\rho_{i}^{o b s}\right) \rho_{i}^{o b s}+\frac{0.1}{N_{I}} \tau^{2}\right.}\right\}+0.01\right] \tag{5.13}
\end{equation*}
$$

where

$$
\begin{array}{ll}
\rho_{i}^{o b s} & \text { is the observed proportion of the landed catch in age-/size-class } i, \\
\rho_{i} & \text { is the model-estimate of the proportion of the landed catch in age-/size-class } i, \\
N_{I} & \text { is the number of age-/size-classes, and } \\
\tau^{2} & \text { is the inverse of minimum of the sample size and } 1000 .
\end{array}
$$

The model-estimates used in Equation 5.13 depend on whether the data are age- or size-composition data:

$$
\begin{array}{ll}
\rho_{t, a}^{g, s, f}=\sum_{a^{\prime \prime}} Y_{a^{\prime}, a}^{s} C_{y, a a^{\prime}}^{L, s, g} / \sum_{a^{\prime}} C_{y, a^{\prime}}^{L, s, g, f} & \text { age-composition data } \\
\rho_{t, l}^{g, s, f}=\sum_{a} X_{a, l}^{g, s} C_{y, a}^{L, s, g, f} / \sum_{a^{\prime}} C_{y, a^{\prime}}^{L, s, g, f} & \text { size-composition data } \tag{5.14}
\end{array}
$$

where $\quad X_{a, l}^{g, s}$ is the probability that an animal of stock $s$, sex $g$ and age $a$ is in size-class $l$,

$$
\begin{equation*}
X_{a, l}^{g, s}=\int_{\bar{L}_{l}-\Delta L / 2}^{\bar{L}_{l}+\Delta L / 2} \frac{1}{\sigma_{a}^{g, s} \ell \sqrt{2 \pi}} e^{-\frac{\left(\ell \mathrm{n} \ell-\ell \ln _{L^{g+s}}^{g, s}\right)^{2}}{2\left(\sigma_{a}^{g, s}\right)^{2}}} d \ell \tag{5.15}
\end{equation*}
$$

$\bar{L}_{l} \quad$ is the midpoint of the $l^{\text {th }}$ length-class,
$\Delta L$ is the width of each length-class,
$\sigma_{a}^{g, s} \quad$ is the coefficient of variation of the length of a fish of stock $s, \operatorname{sex} g$, and age $a$, and
$Y_{a^{\prime \prime}, a}^{s} \quad$ is the age-reading error matrix (the probability that an animal of actual age $a^{\prime}$ is aged to be age $a$ ).

The upper and lower limits of the summation in Equation 5.13 are not necessarily the youngest and oldest age-classes and the smallest and largest size-classes. Rather, age- and size-classes may be pooled to increase sample size.

## Estimates of spawning biomass

The contribution of the estimates of spawning biomass for stock $s$ to the negative of the logarithm of the likelihood function is given by:

$$
\begin{equation*}
\ell n L_{3}=\sum_{y}\left(S_{y}^{s}-B_{y}^{o b s, s}\right)^{2} /\left(2\left(\sigma_{y}^{s}\right)^{2}\right) \tag{5.16}
\end{equation*}
$$

where

$$
B_{y}^{o b s, s} \text { is the estimate of female spawning biomass for stock } s \text { and year } y \text {, and }
$$ $\sigma_{y}^{s} \quad$ is the standard error of $B_{y}^{o b s, s}$.

## Catch-rate series

The contribution of the catch-rate data to the negative of the logarithm of the likelihood function is based on the assumption that fluctuations in catchability are independent and log-normally distributed with a coefficient of variance of $\sigma_{q}^{s, f}$ :

$$
\begin{equation*}
\ell \mathrm{n} L_{4}=\frac{1}{2\left(\sigma_{q}^{s, f}\right)^{2}} \sum_{s} \sum_{f} \sum_{y}\left(\ell \mathrm{n} I_{y}^{s, f}-\ell \mathrm{n}\left[q^{s, f} B_{y}^{e, s, f}\left(1-F_{y}^{s, f} / 2\right)\right]\right)^{2} \tag{5.17}
\end{equation*}
$$

where $\quad q^{s, f} \quad$ is the catchability coefficient for stock $s$ and fleet $f$, $I_{y}^{s, f} \quad$ is the catch-rate index for stock $s$, fleet $f$ and year $y$, and $\sigma_{q}^{s, f} \quad$ is (approximately) the coefficient of variation of the random fluctuations in catchability.

## The penalty functions

## Recruitment residuals

The recruitment residuals are assumed to be normally distributed and correlated among stocks/species, i.e.:

$$
\begin{equation*}
P_{1}=\frac{1}{2} \sum_{y} \underline{\varepsilon}_{y}^{T} \mathbf{V}^{-1} \underline{\varepsilon}_{y} \tag{5.18}
\end{equation*}
$$

where V is a variance-covariance matrix, i.e. for three stocks/species:

$$
\mathbf{V}=\left(\begin{array}{ccc}
\left(\sigma_{R}^{1}\right)^{2} & \tau^{1,2} \sigma_{R}^{1} \sigma_{R}^{2} & \tau^{1,3} \sigma_{R}^{1} \sigma_{R}^{3}  \tag{5.19}\\
\tau^{1,2} \sigma_{R}^{1} \sigma_{R}^{2} & \left(\sigma_{R}^{3}\right)^{2} & \tau^{2,3} \sigma_{R}^{2} \sigma_{R}^{3} \\
\tau^{1,3} \sigma_{R}^{1} \sigma_{R}^{3} & \tau^{2,3} \sigma_{R}^{2} \sigma_{R}^{3} & \left(\sigma_{R}^{3}\right)^{2}
\end{array}\right)
$$

$\tau^{i, j}$ is the apriori correlation between the recruitment residuals for stock/species $i$ and those for stock/species $j$.

## Selectivity deviations

A penalty is placed on the differences for each fleet in the length-at- $50 \%$ selectivity among stocks:

$$
\begin{equation*}
P_{2}=\frac{1}{2 \sigma_{s}^{2}} \sum_{s} \sum_{f}\left(L_{50}^{s, f}-\bar{L}_{50}^{f}\right)^{2} \tag{5.20}
\end{equation*}
$$

where $\quad \bar{L}_{50}^{f}$ is mean (across stocks) length-at-50\%-selectivity for fleet $f$, and $\sigma_{S}^{2}$ is the between-stock variance in the length-at- $50 \%$-selectivity.

## Fishing mortality deviations

A penalty is placed on the relative trend in fishing mortality:

$$
\begin{equation*}
P_{3}=\omega \sum_{f} \sum_{s} \sum_{y}\left(F_{y}^{s, f} / \overline{F^{s, f}}-\overline{\overline{F_{y}^{f}}}\right)^{2} \tag{5.21}
\end{equation*}
$$

where $\frac{\omega}{F^{s, f}}$ is the weight assigned the fishing mortality penalty, $F_{y}^{f} \quad$ is the mean (over years) value of $F_{y}^{s, f} / F^{s, f}$.

## Steepness

The penalty placed on the steepness of the stock recruitment relationship for each stock is based on assumption that the logit of $(h-0.2) / 0.8$ is normally distributed, i.e.

$$
\begin{equation*}
P_{4}=\sum_{s}\left(\frac{\left[\ln \left(\left\{1-h^{s}\right\} /\left\{h^{2}-0.2\right\}\right)-\mu^{s}\right]^{2}}{2\left(\sigma_{h}^{s}\right)^{s}}\right) \tag{5.22}
\end{equation*}
$$

## An example application

The example application is based on seven species (blue grenadier, gemfish, pink ling, spotted warehou, mirror dory, king dory, and offshore ocean perch). Gemfish has been divided into two stocks (eastern and western) for assessment and management purposes (Rowling, 1994) so the example application is based on a total of eight stocks. The five fleets (and the rationale for their selection) are described in Chapter 4. Chapter 4 outline the data on which the example application is based.

## Parameterization

Table 29 lists the parameters of the population dynamics model and how each parameter is treated in the example application. Several of the parameters that could potentially be estimated by fitting the model (e.g. natural mortality) are pre-specified for the purposes of the example application.
The number of selectivity patterns and the form of each selectivity pattern (logistic or normal) was chosen based on preliminary analyses (using AIC to select the most parsimonious, yet adequate, model for selectivity). The selectivity patterns assumed are as follows:

1. Blue grenadier. Two selectivity patterns are estimated: (a) a domed-shaped (three parameter normal) selectivity function for fleets 1-3, and (b) an asymptotic (increasing with length) selectivity function for fleets 4 and 5 .
2. Eastern gemfish. Two selectivity patterns are estimated: (a) an asymptotic (declining with length) selectivity function for fleet 1, and (b) an asymptotic (increasing) selectivity function for fleets 2 and 5.
3. Pink ling. Three selectivity patterns are estimated: (a) an asymptotic (increasing with length) selectivity function for fleets 1 and 2, (b) an asymptotic (increasing with length) selectivity function for fleets 3 and 4 , and (c) an asymptotic (increasing with length) selectivity function for fleet 5.
4. Spotted warehou. Two selectivity patterns are estimated: (a) an asymptotic (increasing with length) selectivity function for fleet 1 , and (b) an asymptotic (increasing with length) selectivity function for fleets 2-5.
5. Mirror dory: A single selectivity pattern (asymptotic logistic function) is estimated and applied to all fleets.
6. Ocean perch: A single selectivity pattern (asymptotic logistic function) is estimated and applied to all fleets.
7. King dory: A single selectivity pattern (asymptotic logistic function) is estimated and applied to all fleets.
8. Western gemfish: A single selectivity pattern (asymptotic logistic function) is estimated and applied to all fleets.
A greater number of selectivity patterns are estimated for the 'data-rich' and 'data-moderate' stocks than for the 'data-poor' stocks. This does not necessarily imply that selectivity is independent of fleet for the 'datapoor' stocks, but rather than the data for these stocks are insufficient to support estimation of more than one selectivity pattern.

## Model fitting

The age- and length-frequency data are pooled at young ages / small sizes and large ages / large sizes. The specifications for these groups were chosen to avoid large numbers of age- and length-classes with virtually no data. The stock-specific plus- and minus-groups are listed along with the effective sample sizes used when weighting the age- and length-frequency data in Table 30. The effective sample sizes in Table 30 were chosen
so that the realized effective sample sizes were similar to those assumed when fitting the model. The value assumed for $\tau^{2}$ for each year, fleet and stock (see Equation 5.13) is calculated using the formula:

$$
\begin{equation*}
\left(\tau^{2}\right)_{y}^{s, f}=1 /\left(\tilde{N}^{s, f} Q_{y}^{s, f} / \bar{Q}^{s, f}\right) \tag{5.23}
\end{equation*}
$$

where $\quad \tilde{N}^{s, f} \quad$ is the effective sample size for stock $s$ and fleet $f$,
$Q_{y}^{s, f} \quad$ is the number of fish measured / aged for stock $s$ and fleet $f$ during year $y$, and
$\bar{Q}^{s, f} \quad$ is average number of fish measured / aged for stock $s$ and fleet $f$.

The coefficients of variation assumed for each of the catch-rate series are listed in Table 31. The example application downweighted the contribution of the discard rate data by 0.001 because the model was not able to mimic these data well.

The results for the example application are based solely on sex-aggregated data (i.e. the sex-specific lengthfrequency information for eastern gemfish and blue grenadier are ignored). Also, even though lengthfrequency data for blue grenadier, pink ling and spotted warehou exist for some years and fleets for which age-length keys are not available, the number of years is very low. These length-frequency data are therefore also ignored when fitting the models

The results for four variants of the assessment model are reported to enable the impact of different assumptions regarding how the modeling framework can be applied to be examined. The factors considered in these four models (Table 32) are: a) whether penalties are placed on between-stock fishing mortality and between-stock selectivity-at-length, b) whether steepness is estimated or pre-specified, and c) whether the catch-rate data for mirror dory and the recent catch-rate series for eastern and western gemfish are included in the analyses. None of the analyses allow for between-stock correlation in the deviations about the stockrecruitment relationship (i.e. $\tau^{i, j}=0$ for all of the analyses of this chapter).

The results of the various models can be compared to address the following questions:
a) What are implications of including between-stock penalties when assessing multiple stocks simultaneously (model A (the 'base-case' model) vs. model B)?
b) What are implications of estimating rather than pre-specifying the value for the steepness of the stock-recruitment relationship (model A vs. model C)?
c) What are implications of ignoring the catch rate data for mirror dory and eastern and western gemfish (model A vs. model D)?

The third of these questions provides a way to determine whether allowing for between-stock constraints can compensate for a lack of catch-rate data (the primary source of information in trends in abundance for SESSF species). Although CPUE series are available for mirror dory and western gemfish, these are omitted from models A-C so that these two species are 'data-poor' in that there is no direct information on relative abundance for them.

The penalties for the models which impose between-stock constraints (models A, C and D) are:

1. Length-at- $50 \%$-selectivity. A constraint is placed in the difference between the length-at- $50 \%$-selectivity for eastern gemfish (fleet 1) and western gemfish (all fleets). The value of $\sigma_{S}^{-}$(see Equation 5.20) is assumed to be equal to 0.05 so that considerable weight is assigned to this penalty.
2. Fishing pattern over time. Two constraints of this type are considered:
a Fishing by fleet 2 on eastern gemfish, pink ling, mirror dory and ocean perch.
b Fishing by fleet 3 on blue grenadier, king dory and western gemfish.
The value of the weight assigned to this penalty ( $\omega$ - see Equation 5.21 ) is assumed to be 10 .

## Quantifying uncertainty

The variances for the estimates of the model parameters and for the other quantities of interest are determined using Bayesian methods. The Metropolis-Hastings variant of the Markov-Chain Monte Carlo (MCMC) algorithm was used to sample 1,800 equally likely parameter vectors from the joint posterior density function. The samples on which inference is based were generated by running 5,000,000 cycles of the MCMC algorithm, discarding the first 500,000 as a burn-in period and selecting every $2,500^{\text {th }}$ parameter vector thereafter. The initial parameter vector was taken to be the vector of maximum posterior density (MPD) estimates.

Preliminary analyses suggested that it was unlikely that the MCMC algorithm would generate samples which adequately reflect the posterior distribution (in the sense of satisfying the statistics and diagnostic plots commonly used to evaluate convergence of the MCMC algorithm (see Chapter 3: Results and Discussion). Therefore, in addition to evaluating uncertainty using Bayesian methods, an approach based on approximing the variance-covariance matrix by inverting the Hessian matrix at the point corresponding to the MPD estimates was applied. Standard deviations (and hence $90 \%$ confidence intervals) were determined for the logarithms of the annual spawning biomass estimates and the annual ratios of spawning biomass to spawning biomass at the start of 1968. The adequacy of these standard deviations depends critically on the extent to which the objective function is quadratic near its minimum.

## Results

## Ability to mimic the data

The evaluation of the ability to mimic the data is based on model D rather the 'base-case' model (model A) or one of the other two models. This is because: a) this model fits to the largest data set (unlike models A-C, the catch-rate data for mirror dory and eastern and western gemfish are included when estimating the values for the model parameters for model D) thereby increasing the chance of conflicts among the various data sources, b) model D includes more penalties than model B), and c) model D has fewer parameters than model C).

Figure 49 plots the observed and model-predicted catch-rate series. The bars in Figure 49 are the 95\% confidence intervals based on the assumed residual standard deviations (Table 33).

[^8]

Figure 49. Observed and model predicted catch-rate series. The model-predictions are based on model D.

It is difficult to evaluate the ability to mimic the catch-rate data based on Figure 49. Figure 50 therefore provides the same information as Figure 49, except that results are shown without the confidence intervals for the data and in terms of observed versus model-predicted catch-rates.


Figure $\mathbf{5 0}$. Observed and model predicted catch-rate series. The model-predictions are based on model D.

Whether the model is able to mimic the data for a catch-rate series adequately can be evaluated by: a) whether the model predictions fall within the confidence intervals for the data (Figure 49), b) whether the relationship between the observed and model-predicted catch-rates falls along the $1: 1$ line (Figure 50), and whether the standard deviation of the residuals is reasonably similar to the assumed residual standard
deviation (Table 33). Model D is generally able to mimic the catch rate data, although there are some noteworthy exceptions.

1. The fits to the catch-rate series for blue grenadier are not as good as expected from the assumed residual standard deviations (the standard deviations of the residuals are quite substantially larger than the assumed standard deviations, with the consequence that some of the model estimates do not intersect the $95 \%$ confidence intervals for the data).
2. The fits to the catch-rate data for eastern gemfish also provide some evidence for an inability to mimic the assessment data. The misfits to the catch-rates for eastern gemfish tend to occur after 1995 (early series) and before 1990 (later series) suggesting that the two catch-rate series are inconsistent to some extent.
3. The model over-predicts the pre-1990 catch-rates for mirror dory; the standardized catch-rate indices suggest stability from 1986 to 2002, but the model indicates a decline in abundance from 1986-95.
Appendix $5.1(\mathrm{a})-(\mathrm{g})$ shows the fits to the catch age-composition data for seven of the eight stocks (no catch age-composition data for king dory are included in the analyses owing to low sample size). Results are shown separately for each combination of stock, year and fleet. The model-predicted effective sample size for each such combination are included in the captions to the figures.
The effective sample sizes for blue grenadier are comparable with the values assumed when fitting the population dynamics model except for the years prior to 1992 (Table 30b; Appendix 5.1a). A major reason for the relatively high model-predicted effective sample sizes is that the fishery has been dominated by the 1994 and 1995 cohorts in recent years. These cohorts have constituted the bulk of the catch (in numbers) since about 1998 (although many animals from these cohorts were discarded by fleets 1-3 during 1996 and 1997). The model is also able to capture some of the other strong cohorts (e.g. that which was spawned in 1986).
The fits to age-composition data for eastern gemfish caught by fleet 1 are relatively poor (model-predicted effective sample sizes between 3 and 16) although the effective sample size assumed when the fitting the model to the data for this fleet is also quite low ( 10 - Table 30b). The catches by this fleet are usually only fish aged 2-4 years and there is little evidence for strong or weak cohorts passing through the catches of this fleet. The fits to age-composition data for fleet 2 are substantially better than to the age-composition data for fleet 1, except after about 1997 when the targeted fishery for eastern gemfish ceased.

The model-predicted effective sample sizes for the fits to the age-composition data for pink ling average close to 10 (Appendix 5.1c) - the value assumed when fitting the model. However, the fits for some fleets / years (e.g. 1989 by fleet 2 ) are very poor. Unlike the case for blue grenadier, and to a lesser extent eastern gemfish, several of the fits to the age-composition data for pink ling show 'runs' of residuals.

Some of the fits to the catch-age composition data for spotted warehou are quite good (e.g. fleet 3, 2001). However, other fits (e.g. to the data for fleet 3 in 1988) are very poor. It should be noted, however, that the sample sizes for some of the early years are very small (e.g. the age-composition data for 1988 is only based 140 otoliths - Table 27). The fits to the data for fleets 3 and 4 for recent years are generally fairly good, which is pleasing because the bulk of the catches of spotted warehou have been taken by fleet 3 since about 1996 (Table 21).
There is very little information on the age-structure of the catch of mirror dory owing primarily to a lack of ageing information. The model does not adequately capture the catch age-composition information for this stock (Appendix 5.1e). The fits to the age-composition data for ocean perch are also poor (Appendix 5.1f) with the model being largely unable to capture the change in the number of animals in the plus-group. The age-composition data for western gemfish do not provide a clear signal regarding trends in fishing mortality and year-class strength with the result that the model fits for western gemfish are also poor (Appendix 5.1g).
Appendix 5.2(a)-(e) shows the fits to the catch length-frequency data for eastern gemfish, mirror dory, ocean perch, king dory and western gemfish. Note that the model is only fitted to the length-frequency information for a given fleet and year if there are no age-composition data for that fleet and year.

The length-frequency data for gemfish caught by fleet 1 (Appendix 5.2a) are very variable with the result that the effective sample sizes are very low for some years. The fits to the length-frequency data for fleet 2 are better, except after 1992 when the fishery changed and stopped focusing on spawning run fish. The model also struggles to mimic the length-frequency data for mirror dory (Appendix 5.2 b ). This is primarily because large fish may or may not be present is some years (e.g. there were many large fish and few small fish in 1987 but exactly the opposite pattern is evident for 2001). The fits to the length-frequency data for ocean perch (Appendix 5.2c) are also very variable. However, unlike the case for the catch age-composition data, the model occasionally under-predicts the number of fish in the plus-group.
The length-frequencies for king dory are highly variable (perhaps due to the impact of small sample size) and, although the model generally captures the dynamics of the plus group, the effective sample sizes are fairly low (Appendix 5.2d). The model generally captures the overall shape of the length-frequency distribution for western gemfish (Appendix 5.2e) although this is definitely not the case for 1992 when the model is unable to mimic even the lengths which form the bulk of the catch.
The model is not able to mimic the changes over time in the discard rate estimates (Figure 51). For example, the model underpredicts the discard rates of blue grenadier between 1995-8. The fits to the discard data for remaining stocks are also poor.

## The selectivity ogives

Figures 52 and 53 show the 13 estimated selectivity patterns (as a function of age - Figure 52, and as a function of length - Figure 53). The results in Figures 52 and 53 are based on model A, the 'base-case' model for the analyses of this chapter. The results Figures 52 and 53 are largely as expected given the catch age- and size-composition data. For example, selectivity for fleets 1-3 for blue grenadier is dome-shaped while fleet 4 is predicted to only capture large animals. The estimated selectivity patterns for pink ling and spotted warehou are worthy of additional comment. Fleet 1 (east coast, non-winter) captures the broadest range of sizes (specifically, this fleet captures the smallest fish), fleets 2-3 capture larger fish, and fleet 5 (non-trawl) captures only large fish.
The selectivity pattern estimated for western gemfish also warrants some comment. A penalty on difference between the length-at- $50 \%$-selectivity for this stock and that of eastern gemfish (fleet 2 ) is included in the objective function. Perhaps as a result of this, the length-at- $50 \%$-selectivity for western gemfish ( 72.3 cm ) is almost identical to that for eastern gemfish caught by fleet $2(71.1 \mathrm{~cm})$. However, there are marked differences in the shape of the selectivity patterns for eastern and western gemfish, with that for western gemfish capturing a wider range of sizes and ages. This is, however, not inconsistent with the length-frequency data for these stocks which are also quite different (Appendix 5.2), presumably due to differences in the nature of the fisheries concerned. Ignoring the penalty on the length-at- $50 \%$-selectivity (model B) leads to a selectivity pattern for western gemfish with a much lower length-at- $50 \%$-selectivity ( 50.7 cm ) and more asymptotic behaviour at large size (Figure 54).


Figure 51. Observed (dots) and model D-predicted (solid lines) discard rates. The vertical bars are 95\% confidence intervals for the data points.


Figure 52. Selectivity as a function of age (males - solid lines; females - dotted lines) for each of the eight stocks considered in the example application. The results in this figure are based on model $A$.


Figure 53. Selectivity as a function of length for each of the eight stocks considered in the example application. The results in this figure are based on model $A$.


Figure 54. Selectivity as a function of length for each of the eight stocks considered in the example application. The results in this figure are based on model B.

## Ability to quantify uncertainty

Even though the MCMC algorithm was conducted based on $5,000,000$ cycles, there is nevertheless evidence for lack of convergence. Table 34 lists the values of the 'single chain Gelman statistic' for the objective function, the estimates of the logarithms of the virgin recruitments, and the selectivity parameters. As noted in Chapter 3: Results and Discussion, a value of 1.05 for 'single chain Gelman statistic' can be considered as evidence for lack of convergence. The results in Table 34 suggest that many of the parameters fail this diagnostic statistic.

Further evidence for lack of convergence of the MCMC algorithm is provided in Figures 55(a)-(d) which show four diagnostic plots (see Chapter 3: results and discussion for details) for the objective function for each of the four models. These plots highlight the very high correlations among elements of the chain. Such correlation could only be removed by increasing the number of cycles to a prohibitive (e.g. 10-100 fold) extent.
One of the reasons for the poor performance of the MCMC algorithm is the high correlation among some of selectivity parameters (Table 35). An attempt to generate a set of parameter vectors from the Bayesian posterior for model A was therefore made in which the values for the selectivity parameters were set equal to the estimates corresponding to the maximum of the posterior distribution. The results of this exercise are summarized in Table 34 (column 'A-2') and in Figure 56. The values for the diagnostics statistics and plots provide no evidence for a lack of convergence of the MCMC algorithm, suggesting that the correlation among the selectivity parameters may well be the reason for the problems evident in Figure 55. The only parameter for which there is some evidence for convergence problems is the virgin recruitment for king dory (species 7 in Figure 56).
Given the relatively little confidence that can be placed in the Bayesian results owing to the evidence for lack of adequate convergence of the MCMC algorithm, uncertainty is quantified by means of asymptotic standard
deviations and confidence intervals and the Bayesian posteriors. Agreement between these two approaches to quantifying uncertainty (both of which are subject to problems) will tend to provide some confidence that the estimates of uncertainty are at least qualitatively reliable.
a) Model A

b) Model B


Figure 55. Diagnostic statistics (see Chapter 3: Results and Discussion for details) for the objective function.

Results are shown for each of the four models.

Information for 'data-rich' species to inform assessments of 'data-poor' species
c) Model C

d) Model D


Figure 55 (continued) Diagnostic statistics (see Chapter 3: Results and Discussion for details) for the objective function.


Figure 56. Diagnostic statistics (see Chapter 3: Results and Discussion for details) for the objective function and the logarithms of the eight virgin recruitments

Results are shown for a variant of model $A$ in which the selectivity parameters are fixed equal to the estimates corresponding to the maximum of the posterior density function.


Figure 56. (continued) Diagnostic statistics (see Chapter 3: Results and Discussion for details) for the objective function and the logarithms of the eight virgin recruitments


Figure 56. (continued) Diagnostic statistics (see Chapter 3: Results and Discussion for details) for the objective function and the logarithms of the eight virgin recruitments

Information for 'data-rich' species to inform assessments of 'data-poor' species

## Detailed assessment results - model A

Figures 57 - 59 summarize the point estimates of the model outputs for model A (the base-case model - see Table 32). Figure 57 plots the time-trajectories of spawning biomass and recruitment for the eight stocks.
The results for blue grenadier, pink ling and spotted warehou are roughly consistent with those expected from the actual assessments for these stocks (there are differences owing, for example, to somewhat different population dynamics models and data set choices). The estimates of spawning biomass for eastern gemfish for the recent past are much more optimistic that would be expected from the most recent assessment of this stock (Punt, 2000). This is due primarily to the lack in model A of an index of relative abundance for eastern gemfish for the years after 1996.
Aspects of the results that are somewhat surprising include that almost all of the estimated recruitments for pink ling and ocean perch are larger than the recruitment in an unfished state. The results also suggest that the western stock of gemfish is currently depleted to well below its pre-fishery level, but that it has been relatively stable at this depleted level for over a decade. Of the other three stocks that have not been assessed using population dynamics models in the past, king dory and ocean perch exhibit noteworthy and continuing declines (although these stocks are both estimated to be above $40 \%$ of the 1968 level at present) while mirror dory is estimated to have been depleted to about $40 \%$ of its pre-fishery abundance since about 1990.


Figure 57. Time-trajectories of spawning biomass and recruitment for the eight stocks considered in the example application.

The results in this figure are based on model A .
Figure 58 plots the time-trajectories of exploitation rate by stock and fleet. The results mimic the patterns expected from the catch data (Table 21) and the changes over time in abundance (Figure 57).


Figure 58. Time-trajectories of exploitation rate by fleet and stock.
The results in this Figure are based on model A.

Figure 59 shows the time-trajectories of the exploitation rates by fleet and stock normalized to the mean exploitation rate by fleet and stock. The results for fleets 1 and (particularly) 4 are fairly variable while the exploitation rates for fleets 2 and 3 (which are penalized to some extent when fitting model A) exhibit what are qualitatively quite similar patterns.

Fleet 1


Fleet 3


Fleet 2


Fleet 4


Figure 59. Time-trajectories of exploitation rate by fleet and stock.
The results in this figure are based on model A.

## Comparison among models

Figures 60 and 61 compare the estimates of spawning biomass for the eight stocks among the four models in Table 32. Results are shown for spawning biomass in absolute terms (Figure 60) and expressed relative to the 1968 spawning biomass (Figure 61). The results anticipated from the structure of the model and the criteria behind how the penalties are implemented would be that the estimates of spawning biomass for all models would be near identical for the data-rich and data-moderate stocks (blue grenadier, eastern gemfish, pink ling, and spotted warehou), but the estimates of spawning biomass for the four data-poor stocks may differ substantially.

The expected outcomes are largely borne out by the results in Figures 60 and 61. Specifically, the estimates of spawning biomass (in absolute and relative terms) for blue grenadier and spotted warehou are essentially indistinguishable among the models. In relative terms, the estimates of spawning biomass for eastern gemfish and pink ling are the same for models A-C (the estimates of spawning biomass for eastern gemfish for model D differ from those for the remaining models because model D includes an additional catch-rate series; Table 32). The estimates of spawning biomass for the three data-poor stocks based on models A and C are very similar. Expressed relative to the spawning biomass in 1968, the estimates of spawning biomass based on model D are also very similar to those based on models A and C. In contrast, the estimates of spawning biomass for ocean perch, king dory and western gemfish based on model B (which ignores the among-stock constraints) are quite different from those based on models $\mathrm{A}, \mathrm{C}$ and D (which include such constraints).


Figure 60. Time-trajectories of spawning biomass by stock for the four models.


Figure 61. Time-trajectories of spawning biomass by stock (expressed as a percentage of the spawning biomass in 1968) for the four models.

The estimates of spawning biomass for western gemfish based on model B increase after 1992 whereas those based on models A, C, and D are stable. In contrast, model B predicts that ocean perch and king dory to be near collapse; models A, C and D suggest instead that these stocks are close to $40 \%$ of the unfished level at present.
Figure 62 illustrates the reason for the difference in results between models $B$ and $A$. Figure 62 shows the time-trajectories of exploitation rate (scaled to the average exploitation rate) for these two models. The exploitation rate on ocean perch by fleet 2 and that on king dory by fleet 3 based on model B both increase substantially towards the end of the time-series (Figure 62a). However, this is not the case for model A (Figure 62b). The rapidly increasing exploitation rates for ocean perch and king dory probably arise to fit some aspect of the length-composition data for these stocks. However, including a penalty on the interspecies patterns in fishing mortality reduces the 'value' of fitting such data.

## a) Model B


b) Model A

Fleet 2


Fleet 3


Figure 62. Time-trajectories of exploitation rate by fleet and stock. The results in this figure are based on models $B$ and $A$.

Figure 63-66 contrast the $90 \%$ confidence intervals ( $90 \%$ probability intervals for the Bayesian analyses) for spawning biomass and spawning biomass expressed relative to that in 1968. Results based on the asymptotic method of variance estimation are shown in Figures 63 and 64 and results based on the sample from the Bayesian posterior distributions are shown in Figures 65 and 66


Figure $63.90 \%$ confidence intervals for spawning biomass based on the asymptotic method of variance estimation.

Results are shown for the eight stocks and the four models (rows and columns respectively).


Figure $64.90 \%$ confidence intervals for spawning biomass expressed relative to that for 1968 based on the asymptotic method of variance estimation.

Results are shown for the eight stocks and the four models (rows and columns respectively).


Figure 65. 90\% probability intervals for spawning biomass based on the samples from the Bayesian posterior distribution.

Results are shown for the eight stocks and the four models (rows and columns respectively).


Figure $66.90 \%$ probability intervals for spawning biomass expressed relative to that for 1968 based on the samples from the Bayesian posterior distribution

Results are shown for the eight stocks and the four models (rows and columns respectively).

The $90 \%$ confidence intervals (based on the asymptotic method of variance estimation) and the $90 \%$ probability intervals (based on the samples from the Bayesian posterior) are generally remarkably similar. This is perhaps surprising given the clear evidence for lack of convergence of the MCMC algorithm (Figure 55; Table 32). The $90 \%$ confidence intervals and the $90 \%$ probability intervals are essentially identical for the four data-rich / data-moderate stocks. The measures of uncertainty for spawning biomass expressed relative to that in 1968 are more similar than those for spawning biomass in absolute terms. The posterior medians for the four data-poor stocks are not the same as the estimates corresponding to maximum of the posterior density function (the posterior modes). For example, the posterior median for the current depletion of the king dory stock is higher than the corresponding posterior mode.
Given the way the objective function is developed, the expectation is that the posterior / confidence intervals for the data-rich and data-moderate stocks would not differ among models whereas the estimates based on models A, C and D would be more precise than those based on model B. The expectation is clearly borne out for mirror dory and western gemfish (the $90 \%$ posterior / confidence intervals are much narrower for models A, C and D than for model B), but it is not the case for ocean perch and king dory. The reason for the latter result is that these stocks are estimated by model B to be virtually extirpated. Unlike the case for the other stocks therefore, the constraint that the population may not be extinct at present comes into play for these stocks. That this is the reason for the difference between the expected result and that in Figures 63 and 64 is supported by the fact that for the Bayesian posteriors (Figures 65 and 66) the posterior medians do not suggest that the king dory stock is virtually extirpated and the posterior intervals are very wide.

## Discussion

The chapter outlines an approach to fisheries stock assessment which is able to assess multiple stocks/species simultaneously accounting for relationships among stocks/species in terms of the length-at- $50 \%$-selectivity, the time-trajectory in fishery exploitation rate by fleet, and the deviations about the stock-recruitment relationships. This approach allows information for data-rich and data-moderate stocks to influence the results of assessments for data-poor stocks.

The advantages of this approach would be expected to be more stability and precision in the results of assessments for data-poor stocks because these results would be less susceptible to being corrupted by the model attempting to fit to some quirk in a noisy data series.

The example application of this chapter is based on five fleets (four trawl fleets and a non-trawl fleet) and eight stocks. The expectations of improved stability and precision are largely borne out - the results of the analyses of this chapter, specifically, the trends in the abundance of ocean perch, king dory and western gemfish differ fairly substantially when among-stock constraints are taken into account. Furthermore, the precision of some of the key outputs from an assessment (the trends in spawning biomass and spawning biomass expressed relative to the unfished level) for the data-poor stocks is much greater when among-stock constraints are taken into account.

One possible negative feature of the approach outlined in this chapter is that the inclusion of among-stock constraints could impact the results for the data-rich and data-moderate stocks. However, there is no evidence for this in Figures 63-66. Another potential disadvantage of the method outlined in this chapter is that poor choices could be made regarding how any among-species constraints are imposed. The changes in the selectivity curve for western gemfish when among-stock constraints are and are not imposed (Figures 53 and 54) perhaps suggests that it is inappropriate to assume that the length-at- $50 \%$-selectivity for western and eastern gemfish should be similar.

Table 29. The parameters of the population dynamics model. Parameters marked by an asterisk are those for which priors could, in principle, be specified.

| Parameter | Treatment |
| :---: | :---: |
| Natural mortality, $M^{g, s}$ | Pre-specified* |
| Virgin recruitment, $\boldsymbol{K}_{0}^{-}$ | Estimated |
| Stock-recruitment steepness, $h^{s}$ | Pre-specified / Estimated * |
| Fecundity-at-age, $I_{a}$ | Pre-specified |
| Proportion spawning, $\mu^{\circ}$ | Pre-specified |
| Recruitment deviations, $\boldsymbol{\varepsilon}_{y}$ | Estimated* |
| Extent of variation in recruitment, $\sigma_{R}^{\circ}$ | Pre-specified* |
| Weight-at-age, $W_{a}^{\prime}$ | Pre-specified |
| Length-at-age, $L_{a}^{\text {a }}$ | Pre-specified |
| Selectivity-at-age, $L_{50}^{s, 1}, L_{95}^{s, 1}, L^{s, t}, \ \Omega_{L}^{u, \prime}, ~ \ \Omega_{R}^{u, \prime}$ | Estimated* |
| Retention probability, $\varphi_{50}{ }^{\prime \prime}{ }^{\prime}, \varphi_{95}{ }^{\prime \prime}$ | Computed from auxiliary information |
| Maximum age, $x$ | Pre-specified |

The values for the pre-specified parameters are listed in Table 26.

Table 30. The plus- and minus groups, and the effective sample sizes for the age- and length-frequency data.
(a) Length-frequency data

| Stock | Minus / plus groups | Fleet |  |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $(\mathrm{cm})$ | 1 | 2 | 3 | 4 | 5 |
| Eastern gemfish | $30 / 110$ | 10 | 25 | - | - | - |
| Mirror dory | $5 / 60$ | 5 | 5 | - | - | - |
| Ocean perch | $5 / 40$ | 5 | 5 | - | - | - |
| King dory | $5 / 50$ | - | - | 5 | - | - |
| Western gemfish | $30 / 100$ | - | - | 5 | - | - |

(b) Age-composition data

| Stock | Minus / plus groups |  |  |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $(\mathrm{yr})$ | Fleet |  |  |  |  |
|  |  | 1 | 2 | 3 | 4 | 5 |
| Blue grenadier | $3 / 15$ | 25 | 25 | 25 | 25 | - |
| Eastern gemfish | $2 / 10$ | 10 | 50 | - | - | - |
| Pink ling | $1 / 13$ | 10 | 10 | 10 | - | 10 |
| Spotted warehou | $1 / 13$ | 10 | 10 | 10 | 10 | - |
| Mirror dory | $2 / 10$ | 5 | 5 | - | - | - |
| Ocean perch | $5 / 20$ | 5 | 5 | - | - | - |
| Western gemfish | $2 / 10$ | - | - | - | 5 | - |

Information for 'data-rich' species to inform assessments of 'data-poor' species

Table 31. The values assumed for the residual standard deviations when fitting the catch-rate indices.

| Stock | Fleet |  |  |  |
| :--- | :---: | :---: | :---: | :---: |
|  | 1 | 2 | 3 | 4 |
| Blue grenadier | - | - | 0.15 | 0.15 |
| Eastern gemfish | - | 0.10 | - | - |
| Pink ling | 0.20 | 0.20 | 0.20 | 0.20 |
| Spotted warehou | 0.20 | 0.20 | 0.20 | 0.20 |
| Mirror dory | - | 0.30 | - | - |
| Western gemfish | - | - | 0.30 | - |

Table 32. The specifications of the four models considered in the analyses of this chapter.

| Model | With all CPUE series | With <br> penalties | Steepness <br> estimated |
| :---: | :---: | :---: | :---: |
| A | No | Yes | No |
| B | No | No | No |
| C | No | Yes | Yes |
| D | Yes | Yes | No |

Table 33. The assumed residual standard deviations for the catch-rate series and the values inferred from the fit of model $D$.

| Stock / fleet | Assumed residual standard <br> deviation | Inferred residual standard deviation |
| :--- | :---: | :---: |
| Blue grenadier - fleet 3 | 0.15 | 0.240 |
| Blue grenadier - fleet 4 | 0.15 | 0.350 |
| Eastern gemfish - fleet 2 | 0.1 | 0.353 |
| Eastern gemfish - fleet 2 | 0.1 | 0.281 |
| Pink ling - fleet 1 | 0.2 | 0.167 |
| Pink ling - fleet 2 | 0.2 | 0.213 |
| Pink ling - fleet 3 | 0.2 | 0.131 |
| Pink ling - fleet 4 | 0.2 | 0.260 |
| Spotted warehou - fleet 1 | 0.2 | 0.522 |
| Spotted warehou - fleet 2 | 0.2 | 0.294 |
| Spotted warehou - fleet 3 | 0.2 | 0.323 |
| Spotted warehou - fleet 4 | 0.2 | 0.172 |
| Western gemfish - fleet 3 | 0.3 | 0.150 |
| Mirror dory- fleet 2 | 0.3 | 0.328 |

Table 34. Values for the single chain Gelman statistic for the objective function, the logarithms of the virgin recruitments, and the selectivity parameters for the four models.

|  | Model |  |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: |
| Quantity | $\mathrm{A}-1$ | $\mathrm{~A}-2$ | B | C | D |
| Objective fn | 1.062 | 0.999 | 1.118 | 1.129 | 1.071 |
| R0-1 | 1.029 | 1.000 | 1.057 | 1.085 | 1.122 |
| R0-2 | 1.197 | 0.999 | 1.129 | 1.097 | 1.121 |
| R0-3 | 1.070 | 1.001 | 1.101 | 1.058 | 1.129 |
| R0-4 | 1.085 | 1.002 | 1.117 | 1.141 | 1.145 |
| R0-5 | 1.308 | 1.004 | 1.062 | 1.096 | 1.553 |
| R0-6 | 1.182 | 1.000 | 1.256 | 1.163 | 1.252 |
| R0-7 | 1.489 | 1.041 | 1.739 | 1.502 | 1.665 |
| R0-8 | 1.235 | 1.001 | 1.477 | 1.101 | 1.248 |
| Select-1 | 1.030 | - | 1.145 | 1.173 | 1.137 |
| Select-2 | 1.099 | - | 1.291 | 1.210 | 1.306 |
| Select-3 | 1.041 | - | 1.084 | 1.197 | 1.117 |
| Select-4 | 1.088 | - | 1.141 | 1.097 | 1.076 |
| Select-5 | 1.060 | - | 1.231 | 1.167 | 1.056 |
| Select-6 | 1.104 | - | 1.145 | 1.225 | 1.335 |
| Select-7 | 1.195 | - | 1.752 | 2.079 | 1.324 |
| Select-8 | 1.101 | - | 1.116 | 1.325 | 1.568 |
| Select-9 | 1.210 | - | 1.089 | 1.446 | 1.147 |
| Select-10 | 1.016 | - | 1.035 | 1.008 | 1.138 |
| Select-11 | 1.089 | - | 1.059 | 1.036 | 1.132 |
| Select-12 | 1.133 | - | 1.095 | 1.227 | 1.115 |
| Select-13 | 1.147 | - | 1.162 | 1.231 | 1.127 |
| Select-14 | 1.223 | - | 1.044 | 1.082 | 1.487 |
| Select-15 | 1.309 | - | 1.631 | 1.101 | 1.244 |
| Select-16 | 1.266 | - | 1.501 | 1.471 | 1.304 |
| Select-17 | 4.114 | - | 2.092 | 1.003 | 1.000 |
| Select-18 | 1.364 | - | 2.075 | 1.441 | 2.858 |
| Select-19 | 2.480 | - | 1.908 | 2.638 | 1.963 |
| Select-20 | 1.102 | - | 1.362 | 1.308 | 1.504 |
| Select-21 | 1.165 | - | 2.519 | 1.214 | 1.136 |
| Select-22 | 1.082 | - | 1.065 | 1.049 | 1.054 |
| Select-23 | 1.090 | - | 1.061 | 1.036 | 1.066 |
| Select-24 | 1.093 | - | 1.087 | 1.259 | 1.240 |
| Select-25 | 1.025 | - | 1.089 | 1.180 | 1.279 |
| Select-26 | 1.035 | - | 1.066 | 1.069 | 1.034 |
| Select-27 | 1.061 | - | 1.065 | 1.054 | 1.069 |
|  |  |  |  |  |  |
|  |  |  |  |  |  |

Two sets of results are shown for model A

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Table 35. Correlation matrix for the 27 selectivity parameters. Correlations that exceed 0.8 in absolute value are indicated in bold-underline.

| 2 | 0.60 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 3 | $\underline{-0.93}$ | -0.55 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 4 | 0.04 | -0.02 | 0.02 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 5 | -0.01 | -0.02 | 0.02 | 0.56 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 6 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 7 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.41 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 8 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.07 | 0.04 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 9 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.04 | 0.02 | 0.62 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 10 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 11 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 12 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 13 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.80 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 14 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.01 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |  |
| 15 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | -0.19 | 1.00 |  |  |  |  |  |  |  |  |  |  |  |
| 16 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | -0.01 | 0.01 | 1.00 |  |  |  |  |  |  |  |  |  |  |
| 17 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 |  |  |  |  |  |  |  |  |  |
| 18 | 0.01 | 0.00 | -0.04 | -0.03 | -0.01 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 |  |  |  |  |  |  |  |  |
| 19 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | -0.45 | 1.00 |  |  |  |  |  |  |  |
| 20 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.07 | 0.04 | $\underline{0.95}$ | 0.59 | 0.00 | 0.00 | 0.00 | 0.00 | 0.01 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 |  |  |  |  |  |  |
| 21 | 0.02 | 0.02 | -0.02 | -0.01 | 0.00 | 0.01 | 0.01 | 0.14 | 0.09 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.02 | 0.00 | 0.15 | 1.00 |  |  |  |  |  |
| 22 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | -0.01 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 |  |  |  |  |
| 23 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.62 | 1.00 |  |  |  |
| 24 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | -0.03 | -0.02 | 1.00 |  |  |
| 25 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | -0.02 | -0.01 | 0.58 | 1.00 |  |
| 26 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 |
| 27 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 1.00 |

Appendix 5.1. Observed (bars) and model D-predicted (solid lines) catch age-compositions.

## a) Blue grenadier



Information for 'data-rich' species to inform assessments of 'data-poor' species
b) Eastern gemfish

c) Pink ling

d) Spotted warehou

e) Mirror dory

f) Ocean perch

g) Western gemfish



Appendix 5.2. Observed (bars) and model D-predicted (solid lines) catch length-frequency information.
a) Eastern gemfish

b) Mirror dory

c) Ocean perch






d) King dory


Information for 'data-rich' species to inform assessments of 'data-poor' species

## e) Western gemfish



# Chapter 6: Simulation Evaluation of Approach to Informing 'Data-Poor' Species Using Information for 'Data-Rich' Species 

## Introduction

The method outlined in Chapter 5 aims to provide improved estimates of management-related quantities such as current biomass, trends in biomass and current depletion for 'data-poor' species using information for 'data-rich' species. However, prior to the adoption and use of any new method of analysis, it is necessary to show that it can perform at least as well as, and preferably better than, alternative approaches. This is particularly the case for methods of analysis (such as that outlined in Chapter 5) that are to be applied to provide the basis for management decisions, given that a poor analysis method can have substantial economic, social, and biological consequences.
It is not valid to evaluate the potential of a new method of analysis by applying it to an actual data set because the correct answer is seldom known for an actual data set (and certainly not for the species / stocks considered in Chapter 5). One way to evaluate a new method is through cross validation: the analysis method is applied to a subset of the data, and the results of that analysis used to make predictions of the remainder of the data set. Unfortunately, cross validation cannot be applied in this instance because the data sets are too small and the evaluation would consequently be based on an unrealistically uninformative data set compared to that to which it is intended to be applied.

Simulation can be used to evaluate elements of a management system. Specifically, simulation can be used to evaluate:
a decision rules (e.g. Butterworth and Bergh, 1993; Cochrane et al., 1998; Butterworth and Punt, 1999; Punt and Smith, 1999; Geromont et al., 1999; Hilborn et al., 2002; Punt et al., 2002b);
b methods of stock assessment (e.g. Kirkwood, 1981; de la Mare, 1986; Punt, 1989, 1990; Pope and Shepherd, 1985; Bence et al., 1993; Patterson and Kirkwood, 1995; NRC, 1998; Sampson and Yin, 1998; Ianelli, 2002; Punt et al., 2002a);
c methods for conducting future projections (e.g. Maunder and Watters, 2003; Punt and Methot, in press);
d empirical indicators of stock status (e.g. Punt et al., 2001b; Fulton et al., 2004); and
e methods of CPUE standardization (e.g. Porch and Scott, 1994; Maunder, 2001).
The advantages of using simulation to evaluate elements of a management system (see Figure 67 for an overview of the steps used when evaluating stock assessment methods) are numerous (e.g. Punt, 1992; Cooke, 1999). These include that there are no consequences to the actual resource of evaluating a poor method and that the true situation is known exactly. However, the validity of the results of a simulation evaluation are only as good as the scenarios represented in the simulations, specifically whether the simulations adequately represent the vagaries associated with monitoring fisheries resources, conducting stock assessments, and implementing decision rules. Therefore, depending on whether the simulations overor under-represent uncertainty, they will tend to over- or under-estimate the utility of the method under investigation. The problem of the simulations not representing reality very well will be reduced to some extent by comparing several alternative methods using the same simulation protocol.


## Figure $\overline{67}$. Flowchart of the approach used to evaluate a method of stock assessment.

The analyses of this chapter compare the default situation of conducting separate assessments of each new stock / species with the approach of Chapter 5 . The success of the method of Chapter 5 can be evaluated by the extent to which:
a including data for 'data-poor' species with those for 'data-rich' species when conducting assessments leads to poorer estimation performance for the 'data-rich' species; and
b including data for 'data-poor' species with those for 'data-rich' species when conducting assessments leads to better estimation performance for the 'data-poor' species.

## The operating model

## Overview

The operating model is set up to mimic roughly the situation to which the assessment method is applied in Chapter 5. It includes four species, one that is 'data-rich' (Species 1), two that are 'data-moderate' (Species 2 and 3) and one that is 'data-poor' (Species 4). Species 1-3 are meant to mimic (roughly) blue grenadier, eastern gemfish and pink ling. The values for the biological (natural mortality, growth, etc.) and technological (e.g. selectivity, probability of discarding) parameters of the operating model for Species 4 are set to those for spotted warehou, but the data generated for Species 4 are much less extensive than is actually the case in reality for spotted warehou. The operating model considers the period 1920-2002. It is projected forward from unexploited equilibrium at the start of 1920 with variable recruitment thereafter. This specification implies that the age-structure of the population will not be in equilibrium at the start of 1968, as is assumed by the method of stock assessment described in Chapter 5.

There are a total of five fleets. Fish of the 'data-rich', 'data-poor' and one of the 'data-moderate' species (Species 3) are caught by all five fleets while fish of the other 'data-moderate' species (Species 2) are only caught by two of the fleets (Fleets 1 and 2).

## Population dynamics model

The population dynamics model component of the operating model is essentially identical to that underlying the stock assessment methods (see Chapter 5: The population dynamics model; Equations 5.1-5.10). A key exception to this is that rather than specifying the historical catches, the historical exploitation rate is instead computed using the equation:

$$
\begin{equation*}
F_{y}^{s, f}=Q^{s} q^{s, f} E_{y}^{f} e^{\varepsilon_{y}^{s, f}-\left(\sigma_{f}^{F}\right)^{2} / 2} \quad \underline{\varepsilon}_{y}^{f} \sim N\left(\underline{0}, \mathbf{V}^{f}\right) \tag{6.1}
\end{equation*}
$$

where $F_{y}^{s, f}$ is the exploitation rate on fully selected animals of species $s$ by fleet $f$ during year $y$, $q^{s, f} \quad$ is the relative catchability of species $s$ by fleet $f$ (Table 36),
$Q^{s} \quad$ is the overall catchability coefficient for species $s$,
$E_{y}^{f} \quad$ is the (relative) effort for fleet $f$ during year $y$,
$\mathbf{V}^{f}$ is the variance-covariance matrix for fleet $f$, i.e.:

$$
V^{t}=\left(\sigma_{y}^{t}\right)^{2}\left|\begin{array}{ccc}
\rho^{t} & 1 & \rho^{t}  \tag{6.2}\\
\text { g species in the } \\
\rho_{\text {residuals }}^{f} & \rho_{\text {about }}^{f}
\end{array}\right|
$$

$\rho^{f} \quad$ is the correlation among species in the $\rho_{\text {residuals about the average exploitation patern }}^{f}{ }^{f}$ and
$\sigma_{y}^{f} \quad$ is the standard deviation of the residuals about the average exploitation pattern.
The time-trajectory of effort by fleet (see Figure 68) captures the situation in which fleets 1 and 2 started first followed later by fleets 3-5. Effort by fleet 2 peaked in the mid-1980s and then declined substantially. Effort by the remaining fleets increased and then remained constant. Roughly speaking, fleets 1 and 2 mimic the trawl fishery off the east coast of Australia while fleets 3-5 mimic the trawl fleets off western Victoria and Tasmania and the non-trawl fishery. The values for $q^{, \prime \prime}$ in Table 36 were chosen to mimic roughly the spatial patterns of fishing-related impact on the four species included in the operating model


## Figure 68. Relative effort for each of the five fleets.

Table 37 lists the baseline values for the parameters of the operating model and the values examined in the tests of sensitivity. Values are not specified in Table 36 for the overall catchability coefficients ( $Q^{\circ}$ in Equation 6.1). The values for these parameters are instead determined so that if the operating model is projected forwards from 1920 to 2003, the depletion in 2003 equals the pre-specified value in Table 37.

Figures 69 and 70 show the relationships between selectivity and age, and between the probability of being

discarded and age in the operating model.
Figure 69.Age-specific selectivity patterns for the four species (solid lines: females; dotted lines: males).


Figure 70. Probability of discarding as a function of age for each of four species.

## Data generation

The data generated by the operating model include landed catches, discard rates, catch-rate-based indices of abundance, catch age-composition data, catch size-composition data, and survey estimates of abundance.

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## Catch data

The landed catches by fleet and year are assumed to be known without error.

## Discard rate

The estimates of discard rate are assumed to be unbiased and log-normally distributed, i.e.:

$$
\begin{equation*}
D_{y}^{o b s, s, f}=\frac{C_{y}^{D, s, t}}{\tilde{C}_{y}^{L, s, f}+\tilde{C}_{y}^{D, s, f}} e^{\delta_{u}^{D, s, f}-\left(\sigma_{d}^{s, f}\right)^{2} / 2} \quad \delta_{y}^{D, s, f} \sim N\left(0 ;\left(\sigma_{d}^{s, f}\right)^{2}\right) \tag{6.3}
\end{equation*}
$$

where $\boldsymbol{\nu}_{y} \quad$ is the observed fraction of the catch (in mass) of species $s$ by fleet $f$ that was discarded during year $y$,
$\ddot{C}_{y}^{D, s, f}$ is total catch (in weight) discarded by fleet, year and sex,
$\tilde{C}_{y}^{L, s, f}$ is total catch (in weight) landed by fleet, year and sex,
$\sigma_{d}^{s, f} \quad$ is (approximately) the coefficient of variation of the discard rates for species $s$ and fleet $f$.
All of the simulations are based on the assumption that estimates of the fraction of the total catch that is discarded are available from 1991-2002 for each fleet. The value of $\sigma_{d}^{\prime \prime \prime}$ is set equal to 0.3 for all years, fleets and species.

## Catch-rate-based indices of abundance

The catch-rate-based indices of abundance are assumed to be related linearly to exploitable biomass, i.e.:

$$
\begin{equation*}
I_{y}^{s, f}=\tilde{q}^{s, f} B_{y}^{e, s, f}\left(1-F_{y}^{s, f} / 2\right) e^{\eta_{y}^{s, f}-\left(\sigma_{y}^{s, f}\right)^{2} / 2} \quad \eta_{y}^{s, f} \sim N\left(0 ;\left(\sigma_{y}^{s, f}\right)^{2}\right) \tag{6.3}
\end{equation*}
$$

where $\tilde{q}^{s, f} \quad$ is the catchability coefficient for species $s$ and fleet $f$,
$I_{y}^{s, f} \quad$ is the catch-rate index for species $s$, fleet $f$ and year $y$, and
$\sigma_{q}^{s, f} \quad$ is (approximately) the coefficient of variation of the random fluctuations in catchability.
Catch-rate indices are available for two of the fleets for the 'data-rich' species, for one of the fleets for one of the 'data-moderate' species (Species 2), and for four fleets for the other 'data-moderate' species (Species 3) no catch-rate indices are available for the 'data-poor' species and no catch-rate indices are available for the fifth fleet. Table 38 lists the specifications for each of the catch-rate series available for assessment purposes.

## Age- and size-composition data

The catch age- and size-composition data are generated by sampling multinomially from the operating model catch-at-age and catch-at-size compositions (see Equation 5.14). The multinomial sample sizes for the age- and size-composition data by fleet and year are listed in Table 39. Age-composition data are only available for Species 1-3 while relatively imprecise size-composition data are available for Species 4 .

## Survey estimates of abundance

Unbiased estimates of spawning biomass are generated for 1994 and 1995 for Species 1. These estimates are assumed to be log-normally distributed with a coefficient of variation of 0.2.

## Specifications for the estimation model

The assessment models applied to the data generated by the operating model are similar to those examined in Chapter 5. The assumptions common to all of the assessment methods considered in this chapter are:
a the rate of natural mortality, $M$, (Table 31) is known for each species;
b maturity as a function of size is known exactly;
c the form of the selectivity pattern for each species (see Figure 69) is known (but the values for the parameters of the selectivity pattern have to be estimated from the data);
d the residual standard deviations for the discard data and the catch-rate data are known exactly, as are the effective sample sizes for the age- and size-composition data;
e the probability of discarding as a function of size (Figure 70) is known exactly;
f there is no ageing error, and ageing error is ignored when fitting the population dynamics model;
$g$ the same plus and minus groups are assumed when fitting the assessment model to the age- and sizecomposition data as was the case when it was fit to the actual data for blue grenadier, eastern gemfish, pink ling and spotted warehou (Table 30);
h recruitment is estimated for a subset of the years included in the operating model (Species 1-1979-2000; Species 2 - 1968-2000; Species 3 - 1986-2000; Species 4 - 1986-2000) - these particular years were chosen to match the years for which age- and size-composition data are available (Table 39); and
i the extent of variation in recruitment is known exactly.

The five assessment models considered in the simulation study are:
a steepness is known exactly and there is no sharing of parameters among species (abbreviation 'Current');
b steepness is known exactly, and there is a prior on changes in exploitation rate among species (species 1-4 for fleet 1 ; species $1-4$ for fleet 2 ; species 1,3 and 4 for fleet 3 ; species 1,3 and 4 for fleet 4 ). A weight of 10 is assigned to this penalty (abbreviation 'with F prior');
c As for the 'with F prior' assessment model, except that a prior is also placed on the difference in the lengths-at- $50 \%$-selectivity for species 1 caught by fleet 4 , species 2 caught by fleet 2 , species 3 caught by all fleets, species 4 caught by all fleets. A weight of 10 is assigned to this penalty (abbreviation 'with F and S priors');
d As for the 'Current' assessment model, except that steepness is estimated (abbreviation 'Current + steepness').
e As for 'with F prior' assessment model, except that steepness is estimated (abbreviation 'With F prior + steepness').

## Performance measures

There are a very large number of potential performance measures for evaluating whether: a) including data for 'data-poor' species with those for 'data-rich' species when conducting assessments leads to poorer estimation performance for the 'data-rich' species; and b) including data for 'data-poor' species with those for 'data-rich' species when conducting assessments leads to better estimation performance for the 'data-poor' species.

The specific performance measures chosen for each species are:
a The time-trajectory for the relative errors (medians and $90 \%$ intervals) for the estimates of spawning biomass.
b The time-trajectory for the medians of the absolute relative errors for the estimates of spawning biomass.
c The time-trajectory for the relative errors (medians and $90 \%$ intervals) for the estimates of the ratios of the spawning biomass each year to the spawning biomass at the start of 1920.

## Information for 'data-rich' species to inform assessments of 'data-poor' species

d The time-trajectory for the medians of the absolute relative errors for the estimates of the ratios of the spawning biomass each year to the spawning biomass at the start of 1920.
e The histogram of relative errors for the estimate of the spawning biomass at the start of 1920, $S_{0}^{s}$.
f The histogram of the relative errors for the estimate of steepness.
g Box and whisker plots for the rel;ative error of:

- The spawning biomass at the start of 1920.
- The spawning biomass at the start of 1986.
- The spawning biomass at the start of 2003.
- The ratio of the spawning biomass at the start of 1986 to that at the start of 1920.
- The ratio of the spawning biomass at the start of 2003 to that at the start of 1920.
- The ratio of the spawning biomass at the start of 1995 to that at the start of 1986.
- The ratio of the spawning biomass at the start of 2003 to that at the start of 1986.

The box and whisker plots provide the primary basis for the evaluation of performance because the results for several assessment models can be displayed on the same plot for a given operating model variant or the results for several operating model variants can be shown for a given assessment model on a single plot. Performance measures a) - f) are used primarily to explore why a particular result arose.
Spawning biomass is the focus for the evaluation because most management objectives for SESSF species are expressed in terms of spawning biomass. Focus is placed on the 1986 and 2003 spawning biomasses because 2003 is the most recent year and 1986 is the first year for which catch-rate data are available for the bulk of the SESSF species for which such information is available (e.g. Table 38). The ratio of the 1995 to the 1986 spawning biomass and the ratio of the 2003 to the 1986 spawning biomass are reported because Punt et al. (2002a) found that the ratio of the spawning biomass in a recent year to $S_{0}^{s}$ can be very poorly determined.

## Results and discussion

## Baseline trial

Each simulation trial consists of 250 simulations. Figure 71 shows box and whisker plots of the relative errors for seven management-related quantities for three assessment model variants for the baseline operating model ( $\sigma_{y}^{f}=0.1 ; \rho^{f}=0.7$; Species 4 depleted to $0.5 S_{0}$ in 2003). The y-axis is square root transformed to better visualize the results. The three assessment model variants in Figure 71 are: 'Current', 'With F prior', and 'With F and S priors'. The relative performances of the three assessment model variants can be compared in terms of whether they lead to improved accuracy (the solid dots closer to the zero line) and increased precision (narrower interquartile ranges). The 'Current', 'With F prior' and 'With F and S priors' variants achieve the lowest absolute median relative errors for 7,12 and 9 respectively of the 28 quantities in Figure 71 while they achieve the lowest interquantile ranges for 10,16 and 2 respectively of these quantities.

The 'With F and S priors' variant is fairly variable (wide interquartile range) and also leads to substantial bias when estimating biomass for the first 'data-moderate' species (Species 2 ) and (to a lesser extent) for the 'datapoor' species (Figure 71). Consequently, this variant of the assessment model is not considered further in the analyses of this chapter.

Comparing the 'With F prior' and 'Current' variants of the assessment model directly, the 'With F prior' variant outperforms the 'Current' variant in terms of both bias and precision (it achieves the lower median relative error for 16 of the 28 quantities and the lower interquartile range for 18 of the 28 quantities; Table 40). Of specific interest may be that the 'With F prior' variant generally outperforms the 'Current' variant for

Species 4 (the 'data-poor' species), although the improvement in performance is frequently not particularly substantial (e.g. the spawning biomass in 1986 for Species 2 and 4).



















Figure 71. Box and whisker plots of the relative errors for seven management-related quantities for three assessment models

[^9]As expected, the interquartile ranges for Species 4 are generally larger than those for Species 1--3 irrespective of the choice of an assessment method, confirming the expectation that while some improved performance may be achieved by placing priors on inter-species differences in parameters, this cannot turn a 'data-poor' species into a 'data-rich' species - that requires actual data. The 'With F prior' variant is more biased and variable for the second 'data-moderate' species (Species 3) than the 'Current' variant (Figure 71), suggesting that the improved estimation performance for Species 4 does come at a price in terms of possible poorer estimation performance for 'data-rich' and 'data-moderate' species.

Figure 72 shows the time-trajectories of relative error ( $1^{\text {st }}$ column: spawning biomass, $2^{\text {nd }}$ column: spawning biomass relative to that in 1920) and histograms of relative error for the spawning biomass in 1920 and steepness (columns 3 and 4). The results in this figure pertain to the baseline operating model and the 'With F prior' variant of the assessment model.

The relative errors for steepness in Figure 72 are all zero because the variant of the stock assessment model concerned 'knows' the correct value of steepness. $S_{0}$ is actually more uncertain for the 'data-rich' species than for the 'data-moderate' and 'data-poor' species. This is likely a reflection of the fact that $\sigma_{r}=1$ for the 'datarich' species and only 0.6 for the remaining species (Table 31). Thus, the assumption that $S_{1968}=S_{0}$ is likely violated to a greater extent for the 'data-rich' species.
As expected from the results of many previous studies, the estimates of spawning biomass relative to that in 1920 are estimated more accurately and precisely than spawning biomass in tonnes. The $90 \%$ intervals of the relative errors of spawning biomass get wider with time. This is unsurprising for the relative measures (because the ratio of spawning biomass to $S_{0}$ is 1 in 1920 by definition), but is still of some concern. It should be noted, however, that the $90 \%$ intervals do not necessarily give a good impression of, for example, the interquartile range.

## Sensitivity tests

The sensitivity tests examine the implications, in terms of the comparison between the 'Current' and 'With F Prior' variants of the assessment model, of: a) the status of Species 4 at the start of 2003, b) the extent of variation and correlation in exploitation rate about that expected from the pre-specified effort patterns, and c) the effective sample size for the size-composition data for Species 4 being 100 rather than 20 . Table 40 summarizes the results of the baseline simulation trial and the sensitivity tests in terms the number of cases (out of 28) in which each model variant achieves the lower median relative error and interquartile range. Results are shown by species and for all species combined.

The 'With F prior' variant outperforms the 'Current' variant to a greater extent in terms of reduced bias for the '2003 depletion $=0.1^{\prime}$, ' $\sigma_{y}^{f}=0.99$ ', and the 'Better data for Species 4 ' sensitivity tests, and in terms of reduced variation for the first of these sensitivity tests. The 'Current' variant never performed better than the 'With F prior' variant for Species 4 (in terms of achieving the greater number of lower median relative errors), and only outperformed the 'with F prior' variant for the $\sigma_{y}^{f}=0$ sensitivity test. Figure 73 compares the 'Current' and 'With F prior' variants for the 'Better data for Species 4' sensitivity test - the only sensitivity test in which the 'Current' variant is more precise on average than the 'With F prior' variant.

One of the key factors driving the results in Figures 71 and 73 and the first eight rows of Table 40 is the fact that recruitment variation occurs prior to 1968 but the assessment models ignores this. Figure 74 and the row 'Deterministic' in Table 40 examine the consequences of there being no recruitment variation before the year assumed by assessment model to be that in which recruitment variation started. Although somewhat unrealistic, the 'Current' variant outperforms the 'With F prior' variant in terms of bias for this sensitivity test. This result is, however, somewhat misleading because both assessment models are biased (Figure 74).


Figure 72.Time-trajectories of relative error (median relative error: solid line; $\mathbf{9 0 \%}$ intervals: dotted lines; median absolute relative error: dashed line) ( $1^{\text {st }}$ column: spawning biomass, $2^{\text {nd }}$ column: spawning biomass relative to that in 1920) and histograms of relative error for the spawning biomass in 1920 and steepness (columns 3 and 4).

The results in this figure pertain to the baseline operating model and the 'With F prior' variant of the assessment model.


Figure 73. Box and whisker plots of the relative errors for seven management-related quantities for two assessment models (A - 'Current', B - 'With F prior') for the sensitivity test in which the size-composition sample size for Species 4 is increased from 20 to 100.


Figure 74. As for Figure 73, except that the results relate to the sensitivity test in which recruitment variability only occurs for recent years.


Figure 75. As for Figure 73, except that steepness is treated as an estimable parameter.


Figure 76. As for Figure 72, except that the assessment model treats steepness as an estimable parameter.


Figure 77. As for Figure 73, except that $\sigma_{r}$ is assumed to be 5 when applying the stock assessment model.

The performance of the 'With F prior' variant deteriorates markedly when steepness is estimated within the assessment (Figures 75 and 76, Table 40 row 'Steepness estimated'). The estimates of steepness are both biased and imprecise (Figure 75) even though a prior is placed on this parameter.

Figure 77 examines the consequences of dropping the penalty on recruitment variation when fitting the model (by setting $\sigma_{r}=5$ ). Performance for the 'data-rich' and 'data-moderate' species does not change much but the estimates of spawning biomass for the 'data-poor' species are much more imprecise than was the case in Figure 71.

## Conclusion

The results of the analyses of this chapter indicate that, in almost all cases, including a prior on the inter-fleet differences in relative exploitation rate is likely to lead to improved estimation performance. Some quantities may, however, be estimated worse when a prior on the exploitation rate is included in the assessment. Overall, however, there does appear to be value in placing priors on the inter-fleet differences in relative exploitation rate.

Table 36. The values for the fleet- and species-specific catchability coefficients.

| Species | Fleets |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 1 | 2 | 3 | 4 | 5 |  |
| 1 | 0.10 | 0.05 | 0.10 | 1.00 | 0.00 |  |
| 2 | 0.25 | 1.00 | 0.00 | 0.00 | 0.00 |  |
| 3 | 0.50 | 0.50 | 0.50 | 0.50 | 1.00 |  |
| 4 | 0.25 | 0.25 | 1.00 | 1.00 | 0.10 |  |

Table 37. Baseline values for the parameters of the operating model and the values considered in the tests of sensitivity.

| Parameter | Baseline value | Sensitivity values |
| :--- | :---: | :---: |
| Depletion in 2003 |  |  |
| Species 1 | 0.6 |  |
| Species 2 | 0.1 |  |
| Species 3 | 0.25 | $0.1 / 0.9$ |
| Species 4 | 0.5 | $0.2 / 0.99$ |
| Correlation in exploitation rate, $\rho^{\prime}$ | 0.7 | $0 / 0.5$ |
| Standard deviation in exploitation rate, $\boldsymbol{U}_{\dot{j}}$ | 0.1 |  |

Table 38. The specifications for the catch-rate series.

| Species | Fleet | First year | Last year | $\sigma_{q}$ |
| :---: | :---: | :---: | :---: | :---: |
| 1 | 3 | 1986 | 2002 | 0.15 |
| 1 | 4 | 1986 | 2002 | 0.15 |
| 2 | 1 | 1973 | 1999 | 0.15 |
| 3 | 1 | 1986 | 2002 | 0.20 |
| 3 | 2 | 1986 | 2002 | 0.20 |
| 3 | 3 | 1986 | 2002 | 0.20 |
| 3 | 4 | 1986 | 2002 | 0.20 |

Table 39. The specifications for the catch age- and size-composition data generated by the operating model.
(a) Catch age-composition data

| Species | Fleet | Year range | Sample size |
| :---: | :---: | :---: | :---: |
| 1 | $1,2,3,4$ | $1979-2002$ | 100 |
| 2 | 1,2 | $1982-1990$ (every 2nd year) | 100 |
| 2 | 1,2 | $1991-2002$ | 100 |
| 3 | $1,2,3,4,5$ | $1987-2002$ | 100 |

(b) Catch size-composition data

| Species | Fleet | Year range | Sample size |
| :---: | :---: | :---: | :---: |
| 1 | $1,2,3,4$ | $1979-2002$ | 100 |
| 2 | 1,2 | $1975-2002$ | 100 |
| 3 | $1,2,3,4,5$ | $1987-2002$ | 100 |
| 4 | $1,2,3,4,5$ | $1987-2002$ | 20 |

Combinations of species and fleet for which age- / size-composition data are not generated are omitted from this table.

Table 40. Comparison of the 'Current' and 'With F prior' variants of the assessment model for the baseline simulation trial and the sensitivity tests.

| Case | Median Relative Error |  |  |  |  | Interquartile Range |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | All | 1 | 2 | 3 | 4 | All | 1 | 2 | 3 | 4 |
| Baseline | 12/16 | $2 / 5$ | $2 / 5$ | 6/1 | $2 / 5$ | 10/18 | $3 / 4$ | $1 / 6$ | 4 / 3 | $2 / 5$ |
| 2003 depletion $=0.1$ | 3/25 | $1 / 6$ | $0 / 7$ | $1 / 6$ | $1 / 6$ | 9/19 | $2 / 5$ | $2 / 5$ | $2 / 5$ | $0 / 7$ |
| 2003 depletion $=0.9$ | 12/16 | $2 / 5$ | $2 / 5$ | $7 / 0$ | $1 / 6$ | 11/17 | 3/4 | 1/6 | 4/3 | $3 / 4$ |
| $\sigma_{y}^{f}=0$ | 11/17 | $2 / 5$ | 1/6 | 4/3 | 4/3 | 12/16 | $3 / 4$ | $3 / 4$ | 5/2 | $1 / 6$ |
| $\sigma_{y}^{f}=0.99$ | 8/20 | $1 / 6$ | $1 / 6$ | $5 / 2$ | $1 / 6$ | 10/18 | $3 / 4$ | $1 / 6$ | $4 / 3$ | $2 / 5$ |
| $\rho^{f}=0.99$ | 12/16 | $2 / 5$ | $2 / 5$ | $7 / 0$ | $1 / 6$ | 11/17 | $3 / 4$ | $1 / 6$ | $4 / 3$ | $3 / 4$ |
| $\rho^{f}=0.2$ | 13/15 | $2 / 5$ | $2 / 5$ | $6 / 1$ | $3 / 4$ | 10/18 | $3 / 4$ | $1 / 6$ | $4 / 3$ | $2 / 5$ |
| Better data for Species 4 | 6/22 | $2 / 5$ | 1/6 | $0 / 7$ | $3 / 4$ | 12/16 | $3 / 4$ | 1/6 | $4 / 3$ | $4 / 3$ |
| Deterministic | 15/13 | 4/3 | $2 / 5$ | $7 / 0$ | $2 / 5$ | 8/20 | 3/4 | $0 / 7$ | $3 / 4$ | $2 / 5$ |
| Steepness estimated | 15/13 | 4/3 | 5/2 | 4/3 | $2 / 5$ | 23/5 | $5 / 2$ | 6/1 | $7 / 0$ | 5/2 |

The values presented are the number of cases (out of 28) in which each model variant achieves the lower median relative error and interquartile range. Results are shown by species and for all species combined.

[^10]
## Chapter 7: Benefits

The benefits of this project will flow directly to the fishers in the trawl and non-trawl sectors of the Southern and Eastern Scalefish and Shark Fishery (SESSF) through an improved basis for providing scientific management advice for 'data-rich' and 'data-poor' species. The assessments of 'data-rich' species will be improved by a stronger scientific basis for the appropriate ranges for consideration in assessments for the steepness of the stock-recruitment relationship and the extent of variation in recruitment. The assessments of 'data-poor' species will be improved because information from assessments of 'data-rich' species can be used formally (rather than informally) to place constraints on model outcomes. Irrespective of whether this tool is used as the basis for actual assessments, it can be used to identify those 'data-poor' species for which additional research and monitoring is most warranted because they are the most likely to be depleted or declining. At present, such species can only be identified robustly using catch-rate indices of abundance and catch curve analyses.
Benefits of this project flow to many of the fisheries managed by AFMA because the priors and relationships in Chapters 2 and 3 can be used for assessments other than for those of SESSF species. The ADMB code for the method outlined in Chapters 3 and 5 can be tailored to other situations fairly straightforwardly.
Additional benefits relate to specific results and approaches taken in the project. The model developed to deal with age-reading errors can be applied to any study or assessment, and provides a basis for determining the relationship between the number of otoliths read several times and the resultant precision and bias of the relationship between ageing error and age. Most CPUE analyses for SESSF species have included an area term in the GLMs to accommodate the impact of differences in availability and abundance among sectors of the fishery. However, the results here show that for many species there is a year*area interaction, indicating that trends in abundance differ among areas. Consequently, future analyses of catch and effort data should, at least initially, be undertaken separately for each area. This will inform future assessments.

# Chapter 8: Further Development 

## Priors for biological parameters

a The ability to conduct meta-analyses and hence identify relationships among biological characteristics depends on the size of the database analyzed. The database considered in the report substantially exceeds that considered by Koopman et al. (2000). However, it is almost certainly the case that additional data exist. An international collaborative effort is needed to expand the existing database and to provide more rigorous peer-review of the basic data. The current database only includes vertebrate fish - future efforts should concentrate on non-teleost fish (in particular sharks and skates), invertebrates, as well as on vertebrates.
b The analyses of this report focused solely on the Beverton-Holt and Ricker forms of the stock-recruitment relationship. Future research should expand the class of stock-recruitment functions to include, for example, the Shepherd form (Shepherd, 1982) and the possibility of depensation at low stock size. Furthermore, consideration should be given to include variables that index changes in climate to assess whether the effects of climate can be detected.
c Inter-species/ stock correlations in recruitment are believed to occur in the SESSF and elsewhere. A metaanalysis should be conducted to quantify the extent of such correlation and how it could be incorporated into stock assessments.

## Data for use in stock assessments

a Multiple copies of the SEF1 database exist. Each of these differs in terms of the errors that have been corrected. A project that leads to a unified, corrected, and easily accessible version of the SEF1 database would facilitate comparison of results among assessment authors and lead to more efficient use of the analyst's time.
b A project should be developed which has the objective of linking the SEF1 records to the SEF2 database. Availability of a linked database would allow different approaches for developing catch lengthfrequencies to be applied.
c Catch-effort standardization methods are not applied in a consistent way among SESSF species. A project that provides a unified approach to catch-effort standardization (including default methods for model selection and default diagnostic statistics) would reduce the potential for arbitriness when conducting catch-effort standardizations. Such a project would also provide guidelines for defining 'fleets' and for defining 'targeted' fishing. In addition, there are statistically significant interactions of catch rate with year for several species. The solution adopted in this study to overcome this problem is to define fleets in terms of a number of spatial zones. However, the general issue of how to develop catch-rate indices in the face of interactions with year warrants further consideration.
d Most of the recent length and age information is stored in the ISMP database. However, the historical information is not stored in a particularly consistent manner. Development of a SEF-wide database that includes all of the data for all of the SEF species would facilitate more efficient use of scarce resources.

## Multi-species stock assessments

a The model of Chapter 5 currently assumes that selectivity-at-length by fleet is time-invariant. Consideration needs to be given to allowing selectivity to change slowly over time, possibly by allowing for a random walk in one of the parameters of the selectivity function.
b Biologically, many of the SESSF species operate on a year which differs from the standard calendar year. For example, assessments of blue warehou are based on a year which ranges from May to April (Punt
and Smith, in press). Changing the framework to accommodate non-standard years may be complicated if the biological years for different species are inconsistent, but could also lead to improved fits to the monitoring data.
c Considerable research has been directed towards defining 'métiers' in European fisheries (e.g. Beaseau, 1988; Biseau and Gondeaux, 1988; Pelletier and Ferraris, 2000). These approaches could be applied to the data for the SESSF to identify fleets for use in stock assessments.
d The estimates of the precision of spawning biomass and spawning biomass expressed relative to the unfished level based on asymptotic and Bayesian methods were very similar even though there was evidence for lack of convergence of the MCMC algorithm. Given that it is not uncommon for there to be problems obtaining convergence of the MCMC algorithm, it would be of interest to determine the value of samples based on a chain that shows evidence for lack on convergence.
e Future evaluation of the method of Chapter 5 should focus on the implications for the performance of the method of: a) greater variation in discard and catch-rates, $b$ ) variation over time in selectivity, including white and red noise; c) ageing error; d) assuming the wrong values for some the pre-specified parameters of the model and / or functional forms; and e) larger and smaller effective sample sizes for the catch ageand size-composition data.

## Chapter 9: Planned Outcomes

The main outcome from this project is the development and testing of an analytical tool that can be applied to the many low value species in the SESSF to help formally assess the status of the stocks. As SESSRAG is currently developing harvest strategies for the fishery, the results of this project provide an improved ability to advise fishery managers on appropriate indicators and reference points for some 'data-poor' quota species that will probably provide better indicators of trend that either CPUE or age-structure alone, because the approach integrates all of the available data. Ultimately it will provide a costs-effective means of undertaking more formal assessments of the many SESSF species that currently have no formal assessment. At the very least, the approach developed here will be of value to identify species that should be the focus of increased data collection and analysis.

Benefits of this project also flow to many of the fisheries managed by AFMA and state agencies because the priors and relationships developed can be used for assessments other than for those of SESSF species.

## Chapter 10: Conclusions


#### Abstract

Stock assessment methods are used to provide estimates of a variety of quantities for use in fisheries management. These quantities include historical and current biomass, the ratio of the current biomass to some historical, target or limit biomass, and the implications of future sequences of catches or levels of fishing effort. These implications often form a key component of the management advice for a fish stock. However, it is, and will likely remain, almost impossible to conduct stock assessments for all of the species for which they are needed unless inferences for 'data-poor' stocks are based in part on information for 'datarich' stocks, and, more generally, on knowledge for other species and stocks. Formally, this information can be included in assessments in the form of 'prior distributions' (or penalty functions).


This report considers two ways to include 'prior distributions' in assessments. Chapters 2 and 3 examine whether the data for 'data-rich' species/stocks can be used to develop prior distributions for some of the key input parameters included in stock assessment models (Objectives 1 and 2). The results in the Chapter 2 identify that the best way to estimate the rate of natural mortality is Hoenig's (1983) equation (or the variant thereof based on the data analysed in this report) while the method developed by Pauly (1980) seems very sensitive to whether data for species with high $M$ are included in the analysis. Chapter 2 identifies several relationships among biological parameters, but the predictive ability of most of these is reasonably weak. The fact that $\ell_{\infty}$ was found to correlate well with several other biological parameters highlights the importance of conducting ageing studies for as many species as possible (even if the sample sizes are only sufficient to estimate the parameters of a growth curve) and / or to conduct a catch-curve analysis.
Chapter 3 analyses data on stock and recruitment to estimate priors for the steepness of the stock-recruitment relationship. These data suggest that steepness for Clupeiformes, Pleuronectiformes, and Gadiformes is higher than that for "other" species. This needs to be accounted for when selecting the data on which to base priors for steepness for SESSF species as only blue grenadier is in this group of "high steepness" fishes. In contract, steepness is notably lower for species not in these three families and this should therefore be expected of most of the species in the SESSF. The results in Chapter 3 found no covariates that explain steepness (although the sample sizes for some of the covariates are quite small) except whether the fish is a Clupeiform, Pleuronectiform, or Gadiformes. The results in Chapter 3 can be used to develop base-case priors for steepness and the extent of variation in recruitment, and the values for these quantities to be used in tests of sensitivity.
The second approach to make use of information for 'data-rich' species when conducting assessments for 'data-poor' species is to conduct assessments of several species ('data-rich' and 'data-poor') simultaneously and to impose penalties on the differences in biological parameters (e.g. steepness, selectivity-at-length) among species/stocks (Objectives 3 to 5 ).
The technical details of such an approach are outlined in Chapter 5 (Objective 3), tested by means of simulation in Chapter 6 (Objective 5), and applied in Chapter 5 to data for eight stocks (seven species) in the SESSF (Objective 4). Note Objective 4 specifies three case studies, but the example application effectively combines four case studies by considering multiple species and multiple areas simultaneously.
The results of the simulations and the example application confirm that there is value in imposing crossstock/species constraints. However, the effects may be quite small, and the consequences, in terms of bias and precision, may be negative for some of the 'data-rich' species. Nevertheless, the assessments for ocean perch and western gemfish are clearly more plausible when among-stock constraints are imposed. The results of the approach in Chapter 5 cannot turn a 'data-poor' stock into a 'data-rich' stock, but will be of value to identify species that should be the focus of increased data collection and analysis. In the context of the example application of this study, the results suggest that the stock of western gemfish may be depleted (although it should be noted that the fishery is at the eastern end of the species' range) while there are declining trends in ocean perch and king dory that warrant further data collection and analysis.

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## Appendix 1: Intellectual Property

No intellectual property has arisen from the project that is likely to lead to significant commercial benefits, patents or licences. Any intellectual property associated with this project will be shared between the Fisheries Research and Development Corporation, PIRVic and the University of Washington.

## Appendix 2: Staff

Mathew Koopman
David C. Smith
André E. Punt

Project Scientist, PIRVic
Research Director, PIRVic
Associate Professor, UW


[^0]:    ${ }^{2}$ Primary Industries Research Victoria
    PO Box 114

[^1]:    ${ }^{1}$ Although reference is made here to 'spawning biomass and recruitment', the data for some of the species are in terms of other units (e.g. biomass of females, egg production). This is, however, not a concern as long as spawning biomass-per-recruit is defined in the correct units.

[^2]:    2 © Otter Software.

[^3]:    Information for 'data-rich' species to inform assessments of 'data-poor' species

[^4]:    ${ }^{3}$ This example effectively combines the three case studies by considering multiple species and multiple areas.

[^5]:    ${ }^{4} x$ is taken to be independent of stock. This assumption can be made without a loss of generality.

[^6]:    Information for 'data-rich' species to inform assessments of 'data-poor' species

[^7]:    ${ }^{5}$ The term 'Selectivity' is used here. In fact, the 'selectivity' is the combined impact of availability and gear selectivity.

[^8]:    Information for 'data-rich' species to inform assessments of 'data-poor' species

[^9]:    (A - 'Current', B - 'With F prior', C - ‘With F and S priors') for the baseline operating model.

[^10]:    Information for 'data-rich' species to inform assessments of 'data-poor' species

