

# Casal2 User Manual

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## 1. Introduction

*KL comment: This paragraph wants re writing Alistair*

CASAL<sup>2</sup> (`casal2`) is a generalised age- or length-structured fish stock assessment model that allows flexibility in specifying population dynamics, parameter estimation and model outputs. CASAL<sup>2</sup> can model population dynamics for an age- or length-structured population using a range of observations, including tagging, relative abundance, and age frequency data. CASAL<sup>2</sup> implements an age-structured population which can have user defined categories (e.g., immature, mature, male, female, predator, prey etc.), and age range.

This manual describes how to use CASAL<sup>2</sup>, including how to run CASAL<sup>2</sup>, how to set up an input configuration file. Further, we describe the population dynamics and estimation methods, and describe how to specify and interpret output.

### 1.1. Version

This document (last modified 2016-05-18) describes CASAL<sup>2</sup> 2016-05-18 (rev. a03d5dd). The CASAL<sup>2</sup> version number is suffixed with a date/time (yyyy-mm-dd) and revision number, giving the revision control system UTC date and revision number for the most recent modification of the source files. User manual updates will usually be issued for each minor version or date release of CASAL<sup>2</sup>, and can be obtained, on request, from the authors.

### 1.2. Citing CASAL<sup>2</sup>

A suitable reference for CASAL<sup>2</sup> and this document is:

S. Rasmussen, I. Doonan, A. Dunn, C. Marsh, K. Large, S. Mormede (2016) CASAL<sup>2</sup> User Manual, 2016-05-18 (rev. a03d5dd). National Institute of Water & Atmospheric Research Ltd. *NIWA Technical Report*. 205 p.

### 1.3. Software license

This program and the accompanying materials are made available under the terms of the licence GNU GPL v2 which accompanies this software (see Section 19).

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### 1.4. System requirements

CASAL<sup>2</sup> is available for most IBM compatible machines running 64-bit Linux and Microsoft Windows operating systems.

Several of CASAL<sup>2</sup>'s tasks are highly computer intensive and a fast processor is recommended. Depending on the model implemented, some of CASAL<sup>2</sup>'s tasks can take a considerable amount of time (minutes to hours), and in extreme cases can even take several days to estimate a model fit. Multi-core machines are necessary when running CASAL<sup>2</sup>.

The program itself requires only a few megabytes of hard-disk space but output files can consume large amounts of disk space. Depending on number and type of user output requests, the output

could range from a few hundred kilobytes to several hundred megabytes. When estimating model fits, several hundred megabytes of RAM may be required, depending on the spatial size of the model, number of categories, and complexity of processes and observations. For extremely large models, several gigabytes of RAM may be required.

### 1.5. Necessary files

For both 64-bit Linux and Microsoft Windows, only the binary file `casal2` or `casal2.exe` is required to run CASAL<sup>2</sup>. No other software is required. We do not compile a version for 32-bit operating systems.

CASAL<sup>2</sup> offers little in the way of post-processing of the output, and a package available that allows tabulation and graphing of model outputs is recommended. We suggest software such as **R** (R Development Core Team 2007) to assist in the post processing of CASAL<sup>2</sup> output. We provide the CASAL2 **R** package for importing the CASAL<sup>2</sup> output into **R** (see Section 14).

### 1.6. Getting help

CASAL<sup>2</sup> is distributed as unsupported software, however we would appreciate being notified of any problems or errors in CASAL<sup>2</sup>. See Section 15.2 for how to report errors, for reporting errors and further information on CASAL<sup>2</sup> contact the development team at [casal2@niwa.co.nz](mailto:casal2@niwa.co.nz).

### 1.7. Technical details

CASAL<sup>2</sup> was compiled on Linux using `gcc`, the C/C++ compiler developed by the GNU Project. The 64-bit Linux version was compiled using `gcc` version 5.2.1 20151010 (Ubuntu Linux). The Microsoft Windows version was compiled using Mingw32 `gcc` (tdm64-1) 5.1.0. The Microsoft Windows installer was built using the Inno Setup 5 application.

CASAL<sup>2</sup> six two minimisers — the first is closely based on the main algorithm of Dennis Jr and Schnabel (1996), and which uses finite difference gradients, and the second is an implementation of the differential evolution solver (Storn and Price, 1995), and based on code by Lester E. Godwin of PushCorp, Inc.. The third and last non auto differential algorithm is the Dlib (King, 2009). ADOLC is an auto differential minimiser more information can be found at Walther et al. (1996), CPPAD is another auto differential minimiser that can be used in CASAL<sup>2</sup> and more information can be found at Wächter and Biegler (2006) and BETADIFF which is a modified version of ADOL-C v1.8.4 by Brian Bull, and is the only auto differential library from the predecessor CASAL.

The random number generator used by CASAL<sup>2</sup> uses an implementation of the Mersenne twister random number generator (Matsumoto and Nishimura, 1998). This, the command line functionality, matrix operations, and a number of other functions use the BOOST C++ library (Version 1.58.0).

Note that the output from CASAL<sup>2</sup> may differ slightly on the different platforms due to different precision arithmetic or other platform dependent implementation issues. The source code for CASAL<sup>2</sup> is available in the windows bundle or on the github repository at <https://github.com/alistairdunn1/CASAL2>

Unit tests of the underlying CASAL<sup>2</sup> code are carried out at build time, using the GOOGLE mock and unit testing framework. The unit test framework aims to cover a significant proportion of the key functionality within the CASAL<sup>2</sup> code base. The unit test code for CASAL<sup>2</sup> is available as a

part of the underlying source code.





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## 2. Model overview

### 2.1. Introduction

CASAL<sup>2</sup> is an age-structured population dynamics model. It implements a statistical catch-at-age population dynamics, using a discrete time-step state-space model that represents a cohort-based population age structure .

CASAL<sup>2</sup> is run from the console window on Microsoft Windows or from a terminal window on Linux. CASAL<sup>2</sup> gets its information from input data files, the main one of which is the *input configuration file*. Commands and subcommands in the input configuration file are used to define the model structure, provide observations, define parameters, and define the outputs (reports) for CASAL<sup>2</sup>. Command line switches tell CASAL<sup>2</sup> the run mode and where to direct its output. See Section sec:running-sam for the details.

We define the model in terms of the *state*. The state consists of two parts, the *partition*, and any *derived quantities* or *derived quantities by cell*. The state will typically change one or more times in every *time-step* of every year, depending on the *processes* defined for each model.

The partition is a representation of the population at an instance in time, and is a matrix of the numbers of individuals within each age, and category. A derived quantity is a cumulative summary of the partition (over all cells) at some point in time. A derived quantity by cell is a cumulative summary of the partition in each of the cells at some point in time. Unlike the partition (which is updated as each new process is applied), each derived quantity records a single value for each year of the model run, and each derived quantity by cell records a layer of values for each year of the model run. Hence, derived quantities build up a vector of values over the model run years. For example, the total number of individuals in a category labelled mature at some point in the annual cycle may be a derived quantity and the total number of individuals in a category labelled mature in each cell of the model at some point in the annual cycle may be a derived quantity by cell. The state is the combination of the partition and any derived quantities or derived quantities by cell at some instance in time. Changes to the state occur by the application of processes. Additions to the vectors of derived quantities occur when a model is requested to add a value to each derived quantity vector.

Running of the model consists of two main parts — first the model state is initialised for a number of iterations (years), then the model runs over a range of predefined years.

The application of processes within each year is controlled by the *annual cycle*. This defines what processes happen in each model year, and in what sequence. Initialisation can be phased, and for each phase, the user need to define the processes that occur in each year, and the order in which they are applied.

For the run years, each year is split up into one or more time-steps (with at least one process occurring in each time-step). You can think of each time-step as representing a particular part of the calendar year, or you can just treat them as an abstract sequence of events.

The division of the year into an arbitrary number of time-steps allows the user to specify the exact order in which processes occur and when observations are evaluated. The user specifies the time-steps, their order, and the processes within each time-step. If more than one process occurs in the same time-step, then they occur in the order that they are specified. Observations are always evaluated at the end of the time-step in which they occur. Hence, time-steps can be used to break processes into groups, and assist in defining the timing of the observations within the annual cycle.

The population structure of CASAL<sup>2</sup> follows the usual population modelling conventions and is similar to those implemented in other population models, for example CASAL (Bull et al., 2012).

The model records the numbers of individuals by age and category (e.g., male, female). In general, cohorts are added via a recruitment event, are aged annually, and are removed from the population via various forms of mortality. The population is assumed to be closed (i.e., no immigration or emigration from the modelled area)

A model is implemented in CASAL<sup>2</sup> using an input configuration file, which is a complete description of the model structure (i.e., spatial and population processes), observations, estimation methods, and reports (outputs) requested. CASAL<sup>2</sup> runs from a console window on Microsoft Windows or from a text terminal on Linux. A model can be either *run*, estimable parameters can be *estimated* or *profiled*, *MCMC* distributions calculated, and these estimates can be used by CASAL<sup>2</sup> as parameters of an operating model to *simulate* observations.

A model in CASAL<sup>2</sup> is specified by an input configuration file, and comprises of four main components. These are the population section (model structure, population dynamics, etc.), the estimation section (methods of estimation and the parameters to be estimated), the observation section (observational data and associated likelihoods), and the report section (printouts and reports from the model). The input configuration file completely describes a model implemented in CASAL<sup>2</sup>. See Sections 8, 9, 10, and 11 for details and specification of CASAL<sup>2</sup>'s command and subcommand syntax within the input configuration file.

## 2.2. The population section

### *check and confirm text*

The population section (Section 4) defines the model of the population dynamics. It describes the model structure (i.e. the population structure), initialisation and run years (model period), population processes (for example, recruitment, migration, and mortality), selectivities, and key population parameters.

## 2.3. The estimation section

### *check and confirm text*

The estimation section (Section 5) specifies the parameters to be estimated, estimation methods, penalties and priors. Estimation is based on an objective function (e.g., negative log posterior). Depending on the run mode, the estimation section is used to specify the methods for finding a point estimate (i.e., the set of parameter values that minimizes the objective function), doing profiles, or MCMC methods and options, etc.

Further, the estimation section specifies the parameters to be estimated within each model run and the estimation methods. The estimation section specifies the choice of estimation method, which model parameters are to be estimated, priors, starting values, and minimiser control values.

Penalties and priors act as constraints on the estimation. They can either encourage or discourage (depending on the specific implementation) parameter estimates that are ‘near’ some value, and hence influence the estimation process. For example, a penalty can be included in the objective function to discourage parameter estimates that lead to models where the recorded catch was unable to be fully taken.

## 2.4. The observation section

### *check and confirm text*

Types of observations, their values, and the associated error structures are defined in the observation

section (Section 6). Observations are data which allow us to make inferences about unknown parameters. The observation section specifies the observations, their errors, likelihoods, and when the observations occur. Examples include relative or absolute abundance indices, proportions-at-age frequencies, tag recapture observations, etc. Estimation uses the observations to find values for each of the estimated parameters so that each observation is ‘close’ (in some mathematical sense) to a corresponding expected value.

## **2.5. The report section**

The report section (Section 7) specifies the model outputs. It defines the quantities and model summaries to be output to external files or to the standard output. While CASAL<sup>2</sup> will provide informational messages to the screen, CASAL<sup>2</sup> will only produce model estimates, population states, and other data as requested by the report section. Note that if no reports are specified, then no output will be produced.



---

### 3. Running CASAL<sup>2</sup>

CASAL<sup>2</sup> is run from the console window (i.e., the DOS command line) on Microsoft Windows or from a terminal window on Linux. CASAL<sup>2</sup> gets its information from input data files, the key one of which is the input configuration file.

The input configuration file is compulsory and defines the model structure, processes, observations, parameters (both the fixed parameters and the parameters to be estimated), and the reports (outputs) requested. The following sections describe how to construct the CASAL<sup>2</sup> configuration file. By convention, the name of the input configuration file ends with the suffix `.cs12`, however, any file name is acceptable.

Other input files can, in some circumstances, be supplied to define the starting point for an estimation or as a point estimate from which to simulate observations.

Simple command line arguments are used to determine the actions or *tasks* of CASAL<sup>2</sup>, i.e., to run a model with a set of parameter values, estimate parameter values (either point estimates or MCMC), project quantities into the future, simulate observations, etc.. Hence, the *command line arguments* define the *task*. For example, `-r` is the *run*, `-e` is the *estimation*, and `-m` is the *MCMC* task. The *command line arguments* are described in Section 3.4.

#### 3.1. Using CASAL<sup>2</sup>

*check and confirm text*

To use CASAL<sup>2</sup>, open a console (i.e. the command prompt) window (Microsoft Windows) or a terminal window (Linux). Navigate to a directory of your choice, where your input configuration files are located. Then type `casal2` with any arguments (see Section 3.4 for the the list of possible arguments). CASAL<sup>2</sup> will print output to the screen and return you to the command prompt when it completes its task. Note that the CASAL<sup>2</sup> executable (binary) must be either in the directory where you run it or somewhere in your systems `PATH`. We are working on a installer which is coming soon. See your operating system documentation for help on identifying or modifying your `PATH`.

#### 3.2. The input configuration file

*check and confirm text*

The input configuration file is made up of four broad sections; the description of the population structure and parameters (the population section), the estimation methods and variables (the estimation section), the observations and their associated likelihoods (the observation section), and the outputs and reports that CASAL<sup>2</sup> will return (the report section). The input configuration file is made up of a number of commands (many with subcommands) which specify various options for each of these components.

The command and subcommand definitions in the input configuration file can be extensive (especially when you have a model that has many observations), and can result in a input configuration file that is long and difficult to navigate. To aid readability and flexibility, we can use the input configuration file command `!include file`. The command causes an external file, *file*, to be read and processed, exactly as if its contents had been inserted in the main input configuration file at that point. The file name must be a complete file name with extension, but can use either a relative or absolute path as part of its name. Note that included files can also contain `!include` commands — but be careful that you do not set up a recursive state. See Section 12 for more detail.

### 3.3. Redirecting standard output

*check and confirm text*

CASAL<sup>2</sup> uses the standard output stream `standard output` to display run-time information. The standard error stream is used by CASAL<sup>2</sup> to output the program exit status and run-time errors. We suggest redirecting both the standard output and standard error into files. With the bash shell (on Linux systems), you can do this using the command structure,

```
(casal2 [arguments] > out) >& err &
```

It may be useful to redirect the standard input, especially if you're using CASAL<sup>2</sup> inside a batch job software, i.e.

```
(casal2 [arguments] > out < /dev/null) >& err &
```

On Microsoft Windows systems, you can redirect to standard output using,

```
casal2 [arguments] > out
```

And, on some Microsoft Windows systems (e.g., Windows7), you can redirect to both standard output and standard error, using the syntax,

```
casal2 [arguments] > out 2> err
```

Note that CASAL<sup>2</sup> outputs a few lines of header information to the output. The header consists of the program name and version, the arguments passed to CASAL<sup>2</sup> from the command line, the date and time that the program was called (derived from the system time), the user name, and the machine name (including the operating system and the process identification number). These can be used to track outputs as well as identifying the version of CASAL<sup>2</sup> used to run the model.

### 3.4. Command line arguments

*check and confirm text*

The call to CASAL<sup>2</sup> is of the following form.:

```
casal2[-c config_file] [task] [options]
```

**-c *config\_file*** Define the input configuration file for CASAL<sup>2</sup>. If omitted, then CASAL<sup>2</sup> looks for a file named `casal2.txt`.

and where *task* is one of;

**-h** Display help (this page).

**-l** Display the reference for the software license (LGPL).

**-v** Display the CASAL<sup>2</sup> version number.

**-r** Run the model once using the parameter values in the input configuration file, or optionally, with the values from the file denoted with the command line argument **-i *file***.

- e** Do a point *estimate* using the values in the input configuration file as the starting point for the parameters to be estimated, or optionally, with the start values from the file denoted with the command line argument `-i file`.
- p** Do a likelihood *profile* using the parameter values in the input configuration file as the starting point, or optionally, with the start values from the file denoted with the command line argument `-i file`.
- m** Do an *MCMC* estimate using the values in the input configuration file as the starting point for the parameters to be estimated, or optionally, with the start values from the file denoted with the command line argument `-i file`.
- f** Project the model *forward* in time using the parameter values in the input configuration file as the starting point for the estimation, or optionally, with the start values from the file denoted with the command line argument `-i file`.
- s number** Simulate the *number* of observation sets using values in the input configuration file as the parameter values, or optionally, with the values for the parameters denoted as estimated from the file with the command line argument `-i file`.

In addition, the following are optional arguments [*options*],

- i file** Input one or more sets of free (estimated) parameter values from *file*. See Section ?? for details about the format of *file*.
- o file** Output a report of the free (estimated) parameter values in a format suitable for `-i file`. See Section ?? for details about the format of *file*.
- g seed** Seed the random number *generator* with *seed*, a positive (long) integer value. Note, if `-g` is not specified, then CASAL<sup>2</sup> will generate a random number seed based on the computer clock time.
- loglevel** arg = {trace, finest, fine, medium} See Section 7.
- tabular** Run with `-r` or `-f` command it will print @report in tabular format. See Section 7.
- single-step** Run with `-r`, this additional option will pause the model and ask the user to specify parameters and their values to use for the next iteration. See Section ??.

### 3.5. Constructing an CASAL<sup>2</sup> input configuration file

The model definition, parameters, observations, and reports are specified in an input configuration file. The population section is described in Section 4 and the population commands in Section 8. Similarly, the estimation section is described in Section 5 and its commands in Section 9, and in Section 7 and Section 11 for the report and report commands.

#### 3.5.1. Commands

CASAL<sup>2</sup> has a range of commands that define the model structure, processes, observations, and how tasks are carried out. There are three types of commands,

1. Commands that have an argument and do not have subcommands (for example, `!include file`)
2. Commands that have a label and subcommands (for example @process must have a label, and has subcommands)

3. Commands that do not have either a label or argument, but have subcommands (for example @model)

Commands that have a label must have a unique label, i.e., the label cannot be used on more than one command of that type. The labels must start with a letter or underscore, can contain letters, underscores, or numbers. Labels must not contain white-space, a full-point ('.'), or other characters that are not letters, numbers, or an underscore. For example,

```
@process BH_Recruitment
or
!include MyModel
```

### 3.5.2. Subcommands

Subcommands in CASAL<sup>2</sup> are for defining options and parameter values for commands. They always take an argument which is one of a specific *type*. The types acceptable for each subcommand are defined in Section 12, and are summarised below.

Like commands (@command), subcommands and their arguments are not order specific — except that all subcommands of a given command must appear before the next @command block. CASAL<sup>2</sup> may report an error if they are not supplied in this way, however, in some circumstances a different order may result in a valid, but unintended set of actions, leading to possible errors in your expected results.

The arguments for a subcommand are either:

<b>switch</b>	true/false
<b>integer</b>	an integer number,
<b>integer vector</b>	a vector of integer numbers,
<b>integer range</b>	a range of integer numbers separated by a hyphen (-), e.g. 1994-1996 2000 is expanded to an integer vector of values 1994 1995 1996 2000),
<b>constant</b>	a real number (i.e. double),
<b>constant vector</b>	a vector of real numbers (i.e. vector of doubles),
<b>estimable</b>	a real number that can be estimated (i.e. estimable double),
<b>estimable vector</b>	a vector of real numbers that can be estimated (i.e. vector of estimable doubles),
<b>string</b>	a categorical (string) value, or
<b>string vector</b>	a vector of categorical values.

Switches are parameters which are either true or false. Enter *true* as true or t, and *false* as false or f.

Integers must be entered as integers (i.e., if year is an integer then use 2008, not 2008.0)

Arguments of type integer vector, integer range, constant vector, estimable vector, or categorical vector contain one or more entries on a row, separated by white space (tabs or spaces).

*Estimable* parameters are those parameters that CASAL<sup>2</sup> can estimate, if requested. If a particular parameter is not being estimated in a particular model run, then it acts as a constant. Within CASAL<sup>2</sup> only estimable parameters can be estimated. And, you have to tell CASAL<sup>2</sup> those that are to be estimated in any particular model. Estimable parameters that are being estimated within a particular model run are called the *estimated parameters*.

### 3.5.3. The command-block format



*check and confirm text*

Each command-block either consists of a single command (starting with the symbol @) and, for most commands, a unique label or an argument. Each command is then followed by its subcommands and their arguments, e.g.,

```
@command, or
@command argument, or
@command label
```

**and then**

```
subcommand argument
subcommand argument
etc.,
```

Blank lines are ignored, as is extra white space (i.e., tabs and spaces) between arguments. But don't put extra white space before a @ character (which must also be the first character on the line), and make sure the file ends with a carriage return.

There is no need to mark the end of a command block. This is automatically recognized by either the end of the file, section, or the start of the next command block (which is marked by the @ on the first character of a line). Note, however, that the *!include* is the only exception to this rule. See Section 12) for details of the use of *!include*.

Note that in the input configuration file, commands, sub-commands, and arguments are not case sensitive. However, labels and variable values are case sensitive. Also note that if you are on a Linux system then external calls to files are case sensitive (i.e., when using *!include file*, the argument *file* will be case sensitive).

Characters used in labels must be alphanumeric and can include underscores (\_). Other characters will result in an error.

**3.5.4. Commenting out lines***check and confirm text*

Text that follows a # on a line are considered to be comments and are ignored. If you want to remove a group of commands or subcommands using #, then comment out all lines in the block, not just the first line.

Alternatively, you can comment out an entire block or section by placing curly brackets around the text that you want to comment out. Put in a { as the first character on the line to start the comment block, then end it with }. All lines (including line breaks) between { and } inclusive are ignored. (These should ideally be the first character on a line. But if not, then the entire line will be treated as part of the comment block.)

```
## This is a comment and will be ignored by CASAL2
@process BH_Recruitment
r0 3000000
{ This
block of code
will also be ignored by CASAL2
}
```

### 3.5.5. Determining parameter names

*check and confirm text*

When CASAL<sup>2</sup> processes a input configuration file, it translates each command and each subcommand into a parameter with a unique name. For commands, this parameter name is simply the command name. For subcommands, the parameter name format is either

`command[label].subcommand` if the command has a label, or  
`command.subcommand` if the command has no label, or  
`command[label].subcommand(i)` if the command has a label and the subcommand arguments are a vector, and we are accessing the *i*th element of that vector.

The unique parameter name is used to reference the parameter when estimating, applying a penalty, or applying a profile. For example, the parameter name of subcommand `r0` of the command `@process` with the label `MyRecruitment` is

```
process[MyRecruitment].r0
```

### 3.6. CASAL<sup>2</sup> exit status values

When CASAL<sup>2</sup> completes its task successfully or errors out gracefully, it returns a single exit status value 'completed' to the standard output. Error messages will be printed to the console and if it is a configuration error. It will print the line number and file the error has been identified at

---

## 4. The population section

### 4.1. Introduction

The population section specifies the model structure, population dynamics, and other associated parameters. It describes the model structure (population structure), defines the population processes (e.g., recruitment, migration, and mortality), selectivities, and model parameters.

The population section consists of several components, including;

- The population structure;
- Model initialisation (i.e., the state of the model at the start of the first year);
- The years over which the model runs (i.e., the start and end years of the model)
- The annual cycle (time-steps and processes that are applied in each time-step);
- The specifications and parameters of the population processes (i.e. processes that add, remove individuals to or from the partition, or shift numbers between ages and categories in the partition);
- Selectivities;
- Parameter values and their definitions;
- Derived quantities, required as parameters for some processes (e.g. spawning stock biomass to resolve the spawner-recruit relationship in a recruitment process).

### 4.2. Population structure

The basic structure of a CASAL<sup>2</sup> population model is defined in terms of an annual cycle, time steps, states, and transitions.

The annual cycle defines what processes happen in each model year, and in what sequence. (In line with the New Zealand fisheries management framework, CASAL<sup>2</sup> runs on an annual cycle rather than, for example, a 6-monthly cycle.)

Each year is split into one or more time steps, with at least one process occurring in each time step. Each time step can be thought of as representing a particular part of the calendar year, or you can just treat them as an abstract sequence of events. **In every time step, there exists a mortality block, this is a block where individuals are removed from the partition. If there are no mortality processes then the mortality block is empty (nothing happens) and occurs at the end of a time step** . maybe the user doesn't need to know this more of a code base statement

The state is the current status of the population, at any given time. The state can change one or more times in every time step of every year. The state object must contain sufficient information to figure out the future course of the fishery (given a model and a complete set of parameters).

There are a number of possible changes in the state, which are called transitions. These include processes such as recruitment, natural mortality, fishing mortality, ageing, migration, tagging events, and maturation.

The division of the year into an arbitrary number of time steps allows the user to specify the exact order in which processes and observations occur. The user needs to specify the time step in which each process occurs. If more than one process occurs in the same time step, they will be applied in the order specified in the `@time_step` block.

The key element of the state is the partition. This is a broadly applicable concept that can be used to describe many different kinds of population model. The partition is simply a breakdown of the total number of fish in the current population into different kinds of fish (Note that the partition records numbers of fish, not biomass). The fish are categorised by various characters. Traditionally these characters have been: length class or age class, sex, maturity, area, stock, tag, and growth-path. However CASAL<sup>2</sup> has no predefined characters in the partition. This is a major extension on CASAL where users can extend on traditional problems, for example incorporating predator, prey, and in the case of some shellfish species a clock(dead) category.

When defining the partition the user must choose:

- whether the partition is subdivided by length class or age class (not both)
- which of the other characters are included in the partition, e.g., the number of areas, stocks, tagging events, or growth paths (if any of these characters are included in the partition).

The resulting partition can be conceptualised as a matrix, where the columns are length or age classes and the rows represent combinations of the other characters. Then the number in each cell of the matrix is the number of fish with the corresponding combination of characters.

#### Craigs attempt

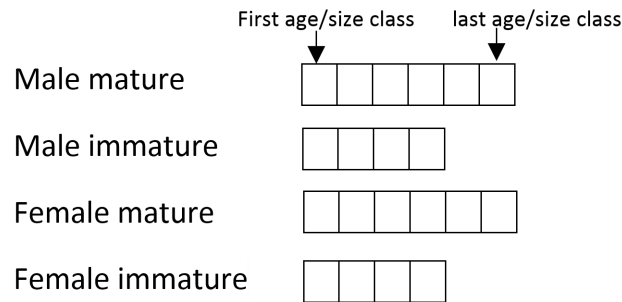
The resulting partition can be conceptualised as a group of vectors, where each category is represented as a vector and the size of the vector is the number of age or length classes, shown in Figure 4.1. Each element in a vector represents the number of individuals for that category in that age or length bin. splitting the partition up into separate categories allows categories to have different age or length structures.

The names and number of categories are user defined, but there must be at least one category defined for a model. The ages are defined as a sequence from  $age_{min}$  to  $age_{max}$ , with the last age optionally a plus group. In order to calculate biomass, the age-length relationship for each category must also be defined. An example of how this is specified for four categories based on sex and area is as follows,

```
@categories
format mature.sex
names spawn.male spawn.female nonspawn.male nonspawn.female
age_lengths male_AL female_AL male_AL female_AL
```

For an example of these ideas, consider a model of a single stock with a spawning and non-spawning fishery. The non-spawning fishery happens over most of the year (say 10 months) in the home area. The mature fish then migrate to the spawning area, where the spawning fishery operates. At the end of spawning, these fish, along with the recruits from the previous year, migrate back to the home area. The modeller decides that fish will be divided in the partition by age, sex, maturity, and area (spawning and home grounds). So the partition has 8 rows (2 sexes (mature or immature) 2 areas) and one column per age class.

For an example of these ideas, consider a model of a single stock with a spawning and non-spawning fishery. The non-spawning fishery happens over most of the year (say 10 months) in the home area. The mature fish then migrate to the spawning area, where the spawning fishery operates. At the end of spawning, these fish, along with the recruits from the previous year, migrate back to the home area. The modeller decides that fish will be divided in the partition by age, sex, maturity, and area (spawning and home grounds). So the partition has 8 rows (2 sexes (mature or immature) 2 areas) and one column per age class.



**Figure 4.1: A visual representation of a partition**

So they define four time steps, labelled 1 through 4. Step 1 includes the non-spawning fishery. Step 2 includes the migration to the spawning area. Step 3 includes the spawning fishery. Step 4 includes recruitment and the migration back to the home area. (In fact, they could have used only 3 time steps, by using a single step in place of their steps 2 and 3. Because the default order of processes within a time step places migrations before fisheries, the processes would still have occurred in the right order.) There are other details to be sorted out, such as the proportion of natural mortality occurring in each time step, but this gives the basic idea.

This structure can be used to implement complex models, with intermingling of separate stocks, with complex migration patterns over multiple areas, and multiple fisheries using different fishing methods and covering different areas and times. Note that there is little point in using a complex structure to model a stock when there are no observations to support that structure. In other words, use a structure for your model that is compatible with the data available.

The model is run from an initial year up to the final(current) year. It can also be run past the final year to make projections things that happen in the future up to the final projection year.

An example, to specify a model with 2 categories (male and female) with ages 1-20 (with the last age a plus group) and an age-length relationship defined with the label `male-growth` and `female-growth`, then the `@model` example from above becomes,

```
@model
start_year
final_year
min_age 1
max_age 20
age_plus_group True
initialisation_phases iphase
time_steps step1 step2
```

### 4.3. The state object and the partition

The key component of the state object is the partition, a matrix of numbers of fish by combinations of characters. The columns can either be age or length classes, the rows are combinations of the following characters:

The key component of the state object is the partition, a group of vectors that store numbers of fish at age or length for a specific category. A category represents a group of fish that have specific attributes, examples of such attributes include life histories and growth paths. Characters in a population that display different attributes and that can make up a category or seperate categories are:

- Sex (male or female);
- Area (any number of areas, named by the user);
- Stock (any number of stocks, named by the user);
- Maturity (immature or mature);
- Growth-path (any number of growth-paths);
- Tag. (any number of tagging events).

A stock is defined as a subpopulation of fish which recruits separately. See Section 4.9 for the treatment of maturity when it is not a character in the partition.

Growth-paths are a feature used to implement some persistence of length at age in an age-based model that uses some length/size data. Each growth-path has its own growth curve, and the length-based model features will consequently have different effects on different growth-paths. So, you need to tell CASAL<sup>2</sup> the following:

- Whether the model is age- or length-based.
- The number and nature of length classes in a length-based model.
- The minimum and maximum age classes in an age-based model.
- Whether there is a plus group.
- The names of all categories and their corresponding growth path labels.
- Whether the partition is divided by sex.
- Whether the partition is divided by maturity.
- Whether the partition has growth-paths, and, if so, how many.
- Whether the partition has multiple stocks, and, if so, how many, and their names.
- Whether the partition has multiple areas, and, if so, how many, and their names.
- Whether the partition includes tagged fish, and, if so, how many, and the names of the tag partitions.

Age classes are always 1 year wide, except that the maximum age group can optionally be a plus group. Users need to choose the minimum and maximum age classes. Length classes are defined by the user, and you need to specify how many length classes there are, the lower bound of each length class, and whether the last length class is a plus group, or if not, what its upper bound is. The relevant parameters are `class_mins` and `plus_group`. The `class_mins` parameter contains the lower bound of each class, and concludes with the upper bound of the last class if it is not a plus group. If, for example, length classes of 30-40, 40-50, 50-60, and 60-70+ cm were desired, you would set `class_mins 30 40 50 60` and `plus_group true`. Whereas if 30-40, 40-50, 50-60, and 60-70 cm were desired, you would set `class_mins 30 40 50 60 70` and `plus_group false`.

CASAL<sup>2</sup> allows categories of the partition to exist for certain years of the model. This is added for computational efficiency, when models contain a large number of categories that do not persist for all model years. Situations where this is beneficial is when a model contains a process that transitions fish into a category that has otherwise not been used, for example tagging events. Excluding categories for certain years saves initialising empty categories. This can be a big time saver if initialisation is run for 50 years and there are many tagging events. (When the ross sea model is up and running % differences would be a nice insert here)

Another important component of the state object in CASAL<sup>2</sup> are derived quantities. This includes quantities such as spawning stock biomasses (SSBs, mid-spawning season biomasses of

spawning fish) for each or a combination of categories. CASAL<sup>2</sup> derives through the command `@derived_quantity`, this is needed if there is a stock-recruitment relationship.

#### 4.4. Time sequences

The time sequence of the model is defined in three parts;

- Initialisation
- Run years
- Projection years

##### 4.4.1. Annual cycle

The annual cycle is implemented as a set of processes that occur, in a user-defined order, within each year. Time-steps are used to break the annual cycle into separate components, and allow observations to be associated with different sets of processes. Any number of processes can occur within each time-step, in any order and can occur multiple times within each time-step. Note that time-steps are not implemented during the initialisation phases (effectively, there is only one time-step), and that the annual cycle in the initialisation phases can be different from that which is applied during the model years.

##### 4.4.2. Initialisation

There are multiple methods to initialise a partition in CASAL<sup>2</sup>. These methods are: iterative, fixed, derived and Cinitial. Model initialisation can occur in several phases, each of which can be a different type. At the end of the initialisation step, CASAL<sup>2</sup> runs through the model years carrying out processes in the order defined in the annual cycle, and can evaluate expected values of observations in order to calculate likelihoods, project forward to determine future states, or simulate observations from the current state.

##### Iterative Initialisation

One of CASAL<sup>2</sup> methods for initialising the initial equilibrium state as an iterative process: a general solution that initialises complex structured models can be difficult to implement using analytic techniques. However, initialising via iteration for a long-lived species with complex transitions can take many iterations and be slow to run. In CASAL<sup>2</sup>, we allow for user-defined multi-phased initialisation using iteration to allow the user to optimize models for speed. Each phase of the initialisation can involve any number of processes. Note that the length of the initialisation period may affect the model outputs, and that a period should be chosen to allow the population state to converge.

In addition, each initialisation process can optionally be stopped early if a user defined convergence criteria is met. For a set of user defined years in the initialisation phase, convergence is defined as met if the proportional absolute summed difference between the the state in year  $t - 1$  and the state in year  $t$  ( $\hat{\lambda}$ ) is less than  $\lambda$  where,

$$\hat{\lambda} = \frac{\sum_i \sum_j |\text{element}(i, j)_t - \text{element}(i, j)_{t-1}|}{\sum_i \sum_j \text{element}(i, j)_t} \quad (4.1)$$

In each initialisation phase, the processes defined for that phase are carried out and used as the starting point for the following phase or, if it is the last phase, then the years that the model is run over. The first phase is always initialised with each element (i.e., each age and category) set at zero. Note that this means that recruitment processes where the numbers of recruits is based on a stock recruitment or density dependant relationship will likely fail if used in the first phase of an initialisation.

The multi-phase iteration allows the user to determine if the initialisation has converged in a particular model run. Here, add an additional initialisation phase for, say, 1 year as the last initialisation phase (with the same processes applied). Then, using the initialisation reports (`@report[label].type=initialisation_partition`), print a copy of the partition just before and just after that phase. If the initialisation has converged to an equilibrium state, then the partition at both these time intervals will be the same.

Hence, for an iterative initialisation you need to define;

- The initialisation phases.
- The number of years in each phase and the processes to apply in each (default is the annual cycle).

### Derived Intitilisation

Derived Intitilisation is an analytical solution to calculate the equilibrium plus group using a geometric series. The benefit of this method is it can be solved in  $\text{max\_age} - \text{min\_age} + 1$  years, so is computationally faster than the iterative initialisation phase. Users should be warned that we have found under some process combinations. This solution does not reach the exact equilibrium partition. We advise to use this method for the computation benefits, but you should always compare to an iterative initialisation to satisfy you have found the equilibrium partition.

### Cinital Intitilisation

This phase can only be applied once a derived or iterative intialisation phases has been implemented. It works off an equilibrium state and uses Cinital factors that can be estimated to shift the initial population away from equilibrium prior to start year. If there is known exploitation before data exists for a population this can be a solution for estimating a non equilibrium population. To apply this method, an observation of age composition data should be provided at start\_year in order to estimate this non equilibrium population. This is implemented in the Southern Blue Whiting assessment [a FAR reference would be nice \(Alistair\)](#)

### Fixed Intitilisation

This is a user defined table that is taken to be the initial partition prior to start year.

### Model years

Following initialisation, the model then runs over a number of user-defined years from (initial\_year to final\_year). For this part of the model, the annual cycle can be broken into separate time-steps, and observations can be associated with the state of the model at the end of any time-step, i.e., likelihoods for particular observations are evaluated, if required, at the end of each time-step.



Processes are carried out in the order specified within each time-step, and can be the same or different to processes in other initialisation phases of the model. The run years define the years over which the model is to run and the annual cycle within each year. The model runs from the start of year `initial` and runs to the end of year `current`. The projection part then extends the run time up to the end of year `final`.

- The time-steps and the processes applied in each
- The initial year (i.e., the model start year)
- The final year (i.e., the model end year)
- The projection final year (i.e., the model projection end year)

#### 4.4.3. Projections

CASAL<sup>2</sup> can project, from a set of parameter estimates, the state of the model into the future. In a projection run, the model is initialised and run through the model years from `initial` to the `final`. Then, the model is re run from `initial` to `projection_final_year`, where any parameter can be projected between this time period. CASAL<sup>2</sup> does not have default projections. Users must specify them using the `@project` blocks. This is important for parameters that are year specific such as year class parameters. If there is no `@project` for these parameters, they will not exist after `final_year` processes that call them will cause nonsensical output. In theory any parameter can be projected forward. The types of `@project` blocks in CASAL<sup>2</sup> are: `constant`, `LogNormal` and `empirical sampling`.

##### Constant

A value is set for the entire model run

##### Empirical Sampling

Parameters that are of type vector or map can be re sampled with replacement between a year range for projected years

##### LogNormal

The randomised parameter are lognormally distributed, with mean 1, and specified standard deviation and autocorrelation on the log-scale.  $YCS_i = \exp(X_i)$ , where  $(X_i)$  are generated as a Gaussian AR(1) process with standard deviation  $\sigma_R$  and mean-0.5  $\sigma_R$  (so that the mean of  $YCS_i$  is 1), and autocorrelation  $\rho$ . Set  $\rho = 0$ , the default, if you don't want autocorrelation. If the randomised parameter are modified by an arbitrary multiplier, then the only change is that parameter will have mean  $\mu$ , where  $\mu$  is the recruitment multiplier.

## 4.5. Population processes

*check and confirm text*

Population processes are those processes that change the population state of individuals. Processes produce changes in the model partition, by adding, removing or moving individuals between ages

or categories. The population processes include recruitment, ageing, mortality events (e.g., natural and exploitation) and category transition processes (i.e., processes that move individuals between categories, while preserving their age structure). See Section 4 for a complete list of available processes.

There are two types of processes, processes that occur across multiple time steps in the annual cycle e.g Natural Mortality and Instantaneous Mortality. There are also processes that only occur within the time step they are defined. Each of these processes is carried out in the user-defined prescribed order when initialising the model, and then for a user-defined order in each year in the annual cycle.

#### 4.5.1. Recruitment

##### *check and confirm text*

Recruitment processes are defined as process that introduces new individuals into the model. CASAL<sup>2</sup> implements two types of recruitment process, constant recruitment and Beverton-Holt recruitment (Beverton and Holt, 1957).

In the recruitment processes, the number of individuals are added to a single age class within the partition, with the amount defined by the type of recruitment process and its function. If more than one category is defined, then the proportion of recruiting individuals to be added to each category is specified by the `proportions` parameter. For example, if recruiting to categories labelled male and female, then you might set the proportions as 0.5 and 0.5 respectively to denote that half of the recruits recruit to the male category and the remaining half to the female category.

For the constant and Beverton-Holt recruitment processes, the number of individuals following recruitment in year  $y$  is,

$$N_{i,j} \leftarrow N_{i,j-1} + p_j(R_y) \quad (4.2)$$

where  $N_{i,j}$  is the numbers in category  $j$  at age  $i$ ,  $p_k$  is the proportion to category  $j$ , and  $R_y$  is the number of recruits for year  $y$ . See below for how  $R_y$  is determined in each of these cases.

#### Constant Recruitment

##### *check and confirm text*

In the constant recruitment process the total number of recruits added each year is  $R_y$ , and is simply  $R_0$ , i.e.

$$R_y = R_0 \quad (4.3)$$

It is equivalent to a Beverton-Holt recruitment process where steepness is set equal to one ( $h = 1$ ).

For example, to specify a constant recruitment process, where individuals are added to the category ‘immature’ at  $age = 1$ , and the number to add is  $R_0 = 5 \times 10^5$ , then the syntax is

```
@process Recruitment
type constant_recruitment
categories immature
proportions 1.0
r0 500000
age 1
```

## Beverton-Holt recruitment

*check and confirm text*

In the Beverton-Holt recruitment process the total number of recruits added each year is  $R_y$ , and is the product of the average recruitment  $R_0$ , the annual year class strength multiplier,  $YCS$ , and the stock-recruit relationship i.e.,

$$R_y = R_0 \times YCS_{y-\text{offset}} \times SR(SSB_{y-\text{offset}}) \quad (4.4)$$

where *offset* is the number of years offset to link the year class with the year of spawning  $y$ , and  $SR$  is the Beverton-Holt stock-recruit relationship parametrised by the steepness  $h$ ,

$$SR(SSB_y) = \frac{SSB_y}{B_0} / \left( 1 - \frac{5h-1}{4h} \left( 1 - \frac{SSB_y}{B_0} \right) \right) \quad (4.5)$$

Note that the Beverton-Holt recruitment process requires a value for  $B_0$  and  $SSB_y$  to resolve the stock-recruitment relationship. Here, a derived quantity (see Section 4.6) must be defined that provides the annual  $SSB_y$  for the recruitment process.  $B_0$  is then defined as the value of the  $SSB$  at the end of one of the initialisation phases. During initialisation the  $YCS$  multipliers are assumed to be equal to one, and recruitment that happens in the initialisation phases that occur before and during the phase when  $B_0$  is determined is assumed to have steepness  $h = 1$  (i.e. in those initialisation phases, recruitment is simply equal to  $R_0$ ). Recruitment in the initialisation phases after the phase where  $B_0$  was determined follow the Beverton-Holt stock-recruit relationship defined above. Recruits are then distributed across cells in proportion to the values in a numeric layer.

For example, assume a Beverton-Holt recruitment process, where individuals are added to the category ‘immature’ at  $\text{age} = 1$ , the number to add is  $R_0 = 5 \times 10^5$ . Then `SSB_Biomass` is a derived quantity that specifies the total spawning stock biomass, with  $B_0$  the value of the derived quantity at the end of the initialisation phase labelled `phase1`. The  $YCS$  are standardised to have mean one in the period 1994 to 2004, and recruits enter into the model two years following spawning. Then the command specification is

```
@process Recruitment
type recruitment_beverton_holt
categories immature
proportions 1.0
r0 500000
steepness 0.75
age 1
ssb SSB_Biomass
standardise_ycs_years 1994-2004
ycs_years 1994 1995 1996 1997 1998 1999 2000 2001 2002 2003 2004 2005 2006
YCS_values 1 1 1 1 1 1 1 1 1 1 1 1 1
```

- If recruitment then ageing then spawning, then `ssb_offset` should equal `min_age + 1`.
- If spawning then ageing then recruitment, then `ssb_offset` should equal `min_age - 1`.
- If any other order is used, then `ssb_offset` should equal `min_age`.

If you have more than one ageing process and a `bevertonholt` recruitment process you will be warned to set your own `ssb_offset` as `CASAL2` will set `ssb_offset` based upon the first ageing process which may be not want the user desires.

### 4.5.2. Ageing

*check and confirm text*

The ageing process simply moves all individuals in the named categories  $i$  to the next age class  $j + 1$ . The ageing process is defined as,

$$\text{element}(i, j) \leftarrow \text{element}(i, j - 1) \quad (4.6)$$

except that in the case of the plus group (if defined),

$$\text{element}(i, \text{age}_{\max}) \leftarrow \text{element}(i, \text{age}_{\max}) + \text{element}(i, \text{age}_{\max-1}). \quad (4.7)$$

For example, to apply ageing to the categories `immature` and `mature`, then the syntax is,

```
@process Ageing
type ageing
categories immature mature
```

Note that ageing is *not* applied by CASAL<sup>2</sup> by default. As with other processes, CASAL<sup>2</sup> will not apply a process unless its defined and specified as a process within the annual cycle. Hence, it is possible to specify a model where a category is not aged. CASAL<sup>2</sup> will not check or otherwise warn if there is a category defined where ageing is not applied.

### 4.5.3. Mortality

*check and confirm text*

Four types of mortality processes are permissible in CASAL<sup>2</sup>, constant rate, event, biomass-event and instantaneous. These processes remove individuals from the partition, either as a rate, as a total number (abundance), as a biomass of individuals or as a mixture of these. CASAL<sup>2</sup> does not implement the Baranov catch equation **YET**. To apply both natural and biomass-event mortality, users can use `mortality_instantaneous`.

**Make a comment about mortality blocks here** for every time step there a mortality block. A mortality block is a number of consecutive mortality processes in a single time step. If no mortality processes occur in a time step then this block defaults to the end of the time step. CASAL<sup>2</sup> will error out if you have multiple mortality events in a time step that are not consecutive events. This mortality block is important for derived quantities see Section 4.6.

#### 4.5.3.1. Constant mortality rate

To specify a constant annual mortality rate ( $M = 0.2$ ) for categories ‘male’ and ‘female’, then,

```
@process NaturalMortality
type mortality_constant_rate
categories male female
selectivities One One
M 0.2 0.2
```

Note that the mortality rate process requires a selectivity. To apply the same mortality rate over all age classes, use a selectivity defined as  $S_j = 1.0$  for all ages  $j$ , e.g.

```
@selectivity One
type constant
c 1
```

#### 4.5.3.2. Event and biomass-event mortality

The event mortality process and biomass mortality processes act in a similar manner, except that they remove a specified abundance (number of individuals) or biomass respectively, rather than applying mortality as a rate. However, the maximum abundance or biomass to remove is constrained by a maximum exploitation rate.

CASAL<sup>2</sup> removes as many individuals or as much biomass as it can while not exceeding the maximum exploitation rate. Event mortality processes require a penalty function to discourage parameter values that do not allow the defined number of individuals to be removed. Here, the model penalises those parameter estimates that result in an insufficient number of individuals in defined categories (after applying selectivities). See Section 5.8 for more information on specifying penalties.

For example, the event mortality applied to user-defined categories  $i$ , with the numbers removed at age  $j$  determined by a selectivity-at-age  $S_j$  is applied as follows:

First, calculate the vulnerable abundance for each category  $i$  in  $1 \dots I$  for ages  $j = 1 \dots J$  that are subject to event mortality,

$$V(i, j) = S(j)N(i, j) \quad (4.8)$$

And hence define the total vulnerable abundance  $V_{total}$  as,

$$V_{total} = \sum_i \sum_j V(i, j) \quad (4.9)$$

Hence the exploitation rate to apply is

$$U = \begin{cases} C/V_{total}, & \text{if } C/V_{total} \leq U_{max} \\ U_{max}, & \text{otherwise} \end{cases} \quad (4.10)$$

And the number removed  $R$  from each age  $j$  in category  $i$  is,

$$R(i, j) = UV(i, j) \quad (4.11)$$

For example, to specify fishing mortality based on catches given for each year, over categories 'immature' and 'mature', with selectivity 'FishingSel' and assuming a maximum possible exploitation rate of 0.7, then the syntax is

```
@process Fishing
type event_mortality
categories immature mature
years 2000 2001 2002 2003
U_max 0.70
selectivities FishingSel FishingSel
penalty event_mortality_penalty
```

#### 4.5.3.3. Instantaneous mortality

The instantaneous mortality process is a combination of natural mortality and event biomass mortality that occurs across multiple time steps. This process applies half the natural mortality, then to apply the mortalities from all the fisheries instantaneously, then to apply the remaining half of the natural mortality.

When Instantaneous mortality is applied the following equations are used.

- An exploitation rate (actually a proportion) is calculated for each fishery, as the catch over the selected-and-retained biomass,

$$U_f = \frac{C_f}{\sum_j \bar{w}_j S_{f,j} n_j e^{-0.5tM_j}}$$

- The fishing pressure associated with fishery  $f$  is defined as the maximum proportion of fish taken from any element of the partition in the area affected by fishery  $f$ ,

$$U_{f,obs} = \max_j (\sum_k S_{k,j} U_k)$$

where the maximum is over all partition elements affected by fishery  $f$ , and the summation is over all fisheries  $k$  which affect the  $j$ th partition element in the same time step as fishery  $f$ .

In most cases the fishing pressure will be equal to the exploitation rate (i.e.,  $U_{f,obs} = U_f$ ). This will not be true only if (a) there is another fishery operating in the same time step as fishery  $f$  and affecting some of the same partition elements, and/or (b) the selectivity  $S_{f,j}$  does not have a maximum value of 1.

There is a maximum fishing pressure limit of  $U_{f,max}$  for each fishery  $f$ . So, no more than proportion  $U_{f,max}$  can be taken from any element of the partition affected by fishery  $f$  in that time step. Clearly  $0 \leq U_{max} \leq 1$ . It is an error if two fisheries which affect the same partition elements in the same time step do not have the same  $U_{max}$ .

For each  $f$ , if  $U_{f,obs} > U_{f,max}$ , then  $U_f$  is multiplied by  $U_{f,max}/U_{f,obs}$  and the fishing pressures are recalculated. In this case the catch actually taken from the population in the model will differ from the specified catch,  $C_f$ .

- The partition is updated using

$$n'_j = n_j \exp(-tM_j) [1 - \sum_f S_{f,j} U_f]$$

An example of the syntax is if we want to apply natural mortality of 0.19 across three time steps on both male and female categories. And we have two fisheries FishingWest FishingEast with there respective catches known for years 1975:1977 in kilograms. These are given in the catches table and information on selectivities, penalties and maximum exploitation rates are given in the fisheries table.

```
@process instant_mort
type mortality_instantaneous
m 0.19
time_step_ratio 0.42 0.25 0.33
selectivities One
categories male female
units kgs
```

```
table catches
year FishingWest FishingEast
1975 80000 111000
1976 152000 336000
1977 74000 1214000
end table
```

```

table fisheries
fishery      category  selectivity u_max  time_step penalty
FishingWest  stock      westFSel   0.7    step1    CatchPenalty
FishingEast  stock      eastFSel   0.7    step1    CatchPenalty
end_table

```

#### 4.5.3.4. Baranov Mortality

Coming Soon!

#### 4.5.4. Transition By Category

This process covers two major processes being, Maturation and Migration. Because CASAL<sup>2</sup> partition is up to the user this type of process should only be used if maturity and/or area are defined in the partition. This processes moves individuals from one category to another, for the case of maturity, this could be moving individuals from immature category to mature category. For migration this could be moving individuals from an area defined category to another.

##### 4.5.4.1. Maturation

Maturation is the process in which immature fish become mature and are moved accordingly in the partition. See Section refsec:maturity-notinpartition for how to treat maturity when it is not a character in the partition.

```

@process Maturation
type category_transition
from male.immature
to male.mature
selectivities MatureSel matureSel
proportions 1 1

```

##### 4.5.4.2. Migration

Migration is the process of moving fish from one area to another. For this to be sensibly applied in CASAL<sup>2</sup> there needs to be a category for the source area and a category for the target area. If two or more migrations are specified in the same time step, then they take place in the order in which they are given.

Aslong as there is one to one category relationship. That is for every source category there is one target category. You can state that a given proportion of these fish migrate (constant across all age or length classes), or provide a selectivity of proportions migrating by age or length class.

$$N_{a,j} = N_{a,i} \times P_i \times S_{a,i} \quad (4.12)$$

where  $N_{a,j}$  is the number of individuals that have moved to category  $j$  from category  $i$  in age  $a$  and  $N_{a,i}$  is the number of individuals in category  $i$ .  $P_i$  is the proportion parameter for category  $i$  and  $S_{a,i}$  is the selectivity at age  $a$  for category  $i$ .

An example, to specify a simple spawning migration of mature males from a western area migrating to an eastern (spawning) area, then the syntax is

```
@process Spawning_migration
type category_transition
from West.males
to East.males
selectivities MatureSel
proportions 1
```

Where `MatureSel` is a selectivity that describes the proportion of age or length classes that are mature and thus move to the eastern area.

### 4.5.5. Tag Release events

Tagging processes can be age or length based processes, where by numbers of fished are moved from an untagged category to a tagged category that the user has defined in the `@Categories` block. Tag release processes can also account for tag induced mortality on individuals. Age based tag release events take a known number of individuals tagged for each age and do a straightforward category transition along with extra mortality. Length based tag release processes are more complicated, as CASAL<sup>2</sup> needs to calculate the age length matrix and exploitation by each length to then move the correct numbers at age based on a length input.

### 4.5.6. Tag Loss

Tag Loss is the process where tags are lost from tagged categories over time from tag failure or getting knocked off. This process is applied as a instantaneous mortality rate that can happen over multiple time steps in the annual cycle. This method assumes when tags are lost that the fish is removed from the partition. All though this seems logically incorrect, we are dealing with such a small number of fish that the impact is minimal and computationally simpler. Note that if your tagging events make up a large proportion of the population you may want to adjust this method. There will be two types of tag loss processes that are termed `single` and `double`. Currently only `single` exists in CASAL<sup>2</sup>. `double` will deal with situations where a tag release process tags individuals with two tags. In which there is another formulae to work out the rate of tag loss.

```
@process Tag_loss
type tag_loss
categories tagged_fish
tag_loss_rate 0.02
time_step_ratio 0.25 0.75
selectivities One
tag_loss_type single
year 1985
```

## 4.6. Derived quantities

Some processes require, as arguments, a population value derived from the population state. These are termed derived quantities. Derived quantities are values, calculated by CASAL<sup>2</sup> as the end of a specified time-step in every year, and hence they have a single value for each year of the model. Derived quantities can be calculated as either an abundance or as a biomass. Abundance derived quantities are simply the count or sum of categories (after applying a selectivity). Biomass derived quantities are similar, except they are a measure of biomass. Derived quantities are also calculated during the initialisation phases, and hence the time-step during each phase must also be specified.



If the initialisation time-steps are not specified, CASAL<sup>2</sup> will calculate the derived quantity during the initialisation phases in every year, at the end of the annual cycle.

Derived quantities are required by some processes, for example the Beverton-Holt recruitment process. The Beverton-Holt recruitment process can require an equilibrium biomass ( $B_0$ ) and annual spawning stock biomass values ( $SSB_y$ ) to resolve the stock-recruit relationship. Here, these would be defined as the abundance or biomass of a part of the population at some point in the annual cycle for selected ages and categories, and would be calculated as a derived quantity.

Derived quantities are associated with a mortality block see section 4.5.3 for more detail on mortality blocks. Users can ask for derived quantities partway through mortality blocks. Currently two methods are implemented in CASAL<sup>2</sup> to interpolate derived quantities part-way through a mortality block, these are `weighted_sum` and `weighted_product`, they are defined as,

- `weighted_sum`: after proportion  $p$  of the mortality block, the partition elements are given by  $n_{p,j} = (1-p)n_j + p'n'_j$
- `weighted_product`: after proportion  $p$  of the mortality block, the partition elements are given by  $n_{p,j} = n_j^{1-p} n_j'^p$

where,  $n_{p,j}$  is the derived quantity at proportion  $p$  of the mortality block for category  $j$ .  $n_j$  is the quantity at the beginning of the mortality block and  $n'_j$  is the quantity at the end of the mortality block.

As an example, to define a biomass derived quantity (say spawning stock biomass, SSB) for a model, evaluated at the end of the first time-step (labelled `step_one`), over all 'mature' male and female categories and halfway through the mortality block using the `weighted_sum` method, we would use the syntax,

```
@derived_quantity SSB
type biomass
time_step step_one
categories mature.male mature.female
selectivities One
time_step_proportion 0.5
time_step_proportion_method weighted_sum
```

## 4.7. Age-length relationship

The age-length relationship defines the length at age (and the weight at length, see Section ??) of individuals at age/category within the model. There are three length-age relationships available in CASAL<sup>2</sup>. The first is the naive no relationship (where each individual has length 1 irrespective of age). The second and third are the von-Bertalanffy and Schnute relationships respectively. The length-at-age relationship is used to determine the length frequency, given age, and then with the length-weight relationship, a weight-at-age of individuals within an age/category.

The three age-length relationships are,

None: where the length of each individual is exactly 1 for all ages, in which case the `none` length-weight relationship must also be used.

von Bertalanffy: where length at age is defined as,

$$\bar{s}(age) = L_{\infty} (1 - \exp(-k(age - t_0))) \quad (4.13)$$

Schnute: where length at age is defined as,

$$\bar{s}(\text{age}) = \begin{cases} \left[ y_1^b + (y_2^b - y_1^b) \frac{1 - \exp(-a(\text{age} - \tau_1))}{1 - \exp(-a(\tau_2 - \tau_1))} \right]^{1/b}, & \text{if } a \neq 0 \text{ and } b \neq 0 \\ y_1 \exp \left[ \ln(y_2/y_1) \frac{1 - \exp(-a(\text{age} - \tau_1))}{1 - \exp(-a(\tau_2 - \tau_1))} \right], & \text{if } a \neq 0 \text{ and } b = 0 \\ \left[ y_1^b + (y_2^b - y_1^b) \frac{\text{age} - \tau_1}{\tau_2 - \tau_1} \right]^{1/b}, & \text{if } a = 0 \text{ and } b \neq 0 \\ y_1 \exp \left[ \ln(y_2/y_1) \frac{\text{age} - \tau_1}{\tau_2 - \tau_1} \right], & \text{if } a = 0 \text{ and } b = 0 \end{cases} \quad (4.14)$$

The von Bertalanffy curve is parameterised by  $L_\infty$ ,  $k$ , and  $t_0$ ; the Schnute curve (Schnute, 1981) by  $y_1$  and  $y_2$ , which are the mean lengths at reference ages  $\tau_1$  and  $\tau_2$ , and  $a$  and  $b$  (when  $b = 1$ , this reduces to the von Bertalanffy with  $k = a$ ).

When defining length-at-age in CASAL<sup>2</sup>, you must also define a length-weight relationship (see Section 4.7 below).

### Calculation of length-at-age (in an age-based model)

#### Interpolation of length-at-age

#### Size-weight relationship

There are two length-weight relationships available in CASAL<sup>2</sup>. The first is the naive no relationship. Here, the weight of an individual, regardless of length, is always 1. The second is the basic relationship.

The two length-weight relationships are,

- None: The length-weight relationship where

$$\text{mean weight} = 1 \quad (4.15)$$

- Basic: The length-weight relationship where the mean weight  $w$  of an individual of length  $l$  is

$$w = al^b \quad (4.16)$$

Note that if a distribution of length-at-age is specified, then the mean weight is calculated over the distribution of lengths, and is

$$w = (al^b)(1 + cv^2)^{\frac{b(b-1)}{2}} \quad (4.17)$$

where the  $cv$  is the c.v. of lengths-at-age. This adjustment is exact for lognormal distributions, and a close approximation for normal distributions if the c.v. is not large (Bull et al., 2012).

Be careful about the scale of  $a$  — this can easily be specified incorrectly. If the catch is in tonnes and the growth curve in centimetres, then  $a$  should be on the right scale to convert a length in centimetres to a weight in tonnes. Note that there are reports available that can be used to help check that the units specified are plausible (see Section 7).

**Calculation of mean weight**

**4.8. Weightless model**

**4.9. Maturity, in models without maturing in the partition**

#### 4.10. Selectivities

A selectivity is a function that can have a different value for each age class. Selectivities are used throughout CASAL<sup>2</sup> to interpret observations (Section 5) or to modify the effects of processes on each age class (Section 4). CASAL<sup>2</sup> implements a number of different parametric forms, including logistic, knife edge, and double normal selectivities. Selectivities are defined in there own command block (@selectivity), where the unique label is used by observations or processes to identify which selectivity to apply.

Selectivities are indexed by age, with indices from `min_age` to `max_age`. For example, you might have an age-based selectivity that was logistic with 50% selected at age 5 and 95% selected at age 7. This would be defined by the `type=logistic` with parameters  $a_{50} = 5$  and  $a_{t095} = (7 - 5) = 2$ . Then the value of the selectivity at age  $x = 7$  is 0.95 and the selectivity at  $x = 3$  is 0.05. Note selectivities can be length based, However Caution, more testing is needed for this functionality.

Note that the function values for some choices of parameters for some selectivities can result in an computer numeric overflow error (i.e., the number calculated from parameter values is either too large or too small to be represented in computer memory). CASAL<sup>2</sup> implements range checks on some parameters to test for a possible numeric overflow error before attempting to calculate function values. For example, the logistic selectivity is implemented such that if  $(a_{50} - x)/a_{t095} > 5$  then the value of the selectivity at  $x = 0$ , i.e., for  $a_{50} = 5$ ,  $a_{t095} = 0.1$ , then the value of the selectivity at  $x = 1$ , without range checking would be  $7.1 \times 10^{-52}$ . With range checking, that value is 0 (as  $(a_{50}x)/a_{t095} = 40 > 5$ ).

The available selectivities are;

- Constant
- Knife-edge
- All values
- All values bounded
- Increasing
- Logistic
- Inverse logistic
- Logistic producing
- Double normal
- Double exponential
- Cubic spline (Not yet implemented)

The available selectivities are described below.

##### 4.10.1. constant

$$f(x) = C \tag{4.18}$$

The constant selectivity has the estimable parameter  $C$ .

**4.10.2. knife\_edge**

$$f(x) = \begin{cases} 0, & \text{if } x < E \\ \alpha, & \text{if } x \geq E \end{cases} \quad (4.19)$$

The knife-edge ogive has the estimable parameter  $E$  and a scaling parameter  $\alpha$ , where the default value of  $\alpha = 1$

**4.10.3. all\_values**

$$f(x) = V_x \quad (4.20)$$

The all-values selectivity has estimable parameters  $V_{low}, V_{low+1} \dots V_{high}$ . Here, you need to provide the selectivity value for each age class.

**4.10.4. all\_values\_bounded**

$$f(x) = \begin{cases} 0, & \text{if } x < L \\ V_x, & \text{if } L \leq x \leq H \\ V_H, & \text{if } x > H \end{cases} \quad (4.21)$$

The all-values-bounded selectivity has non-estimable parameters  $L$  and  $H$ . The estimable parameters are  $V_L, V_{L+1} \dots V_H$ . Here, you need to provide an selectivity value for each age class from  $L \dots H$ .

**4.10.5. increasing**

$$f(x) = \begin{cases} 0, & \text{if } x < L \\ f(x-1) + \pi_x(\alpha - f(x-1)), & \text{if } L \leq x \leq H \\ f(\alpha), & \text{if } x \geq H \end{cases} \quad (4.22)$$

The increasing ogive has non-estimable parameters  $L$  and  $H$ . The estimable parameters are  $\pi_L, \pi_{L+1} \dots \pi_H$  (but if these are estimated, they should always be constrained to be between 0 and 1).  $\alpha$  is a scaling parameter, with default value of  $\alpha = 1$ . Note that the increasing ogive is similar to the all-values-bounded ogive, but is constrained to be non-decreasing.

**4.10.6. logistic**

$$f(x) = \alpha / [1 + 19^{(a_{50}-x)/a_{t095}}] \quad (4.23)$$

The logistic selectivity has estimable parameters  $a_{50}$  and  $a_{t095}$ .  $\alpha$  is a scaling parameter, with default value of  $\alpha = 1$ . The logistic selectivity takes values  $0.5\alpha$  at  $x = a_{50}$  and  $0.95\alpha$  at  $x = a_{50} + a_{t095}$ .

**4.10.7. inverse\_logistic**

$$f(x) = \alpha - \alpha / [1 + 19^{(a_{50}-x)/a_{t095}}] \quad (4.24)$$

The inverse logistic selectivity has estimable parameters  $a_{50}$  and  $a_{t095}$ .  $\alpha$  is a scaling parameter, with default value of  $\alpha = 1$ . The logistic selectivity takes values  $0.5\alpha$  at  $x = a_{50}$  and  $0.95\alpha$  at  $x = a_{50} - a_{t095}$ .

**4.10.8. logistic\_producing**

$$f(x) = \begin{cases} 0, & \text{if } x < L \\ \lambda(L), & \text{if } x = L \\ (\lambda(x) - \lambda(x-1)) / (1 - \lambda(x-1)), & \text{if } L < x < H \\ 1, & \text{if } x \geq H \end{cases} \quad (4.25)$$

The logistic-producing selectivity has the non-estimable parameters  $L$  and  $H$ , and has estimable parameters  $a_{50}$  and  $a_{t095}$ .  $\alpha$  is a scaling parameter, with default value of  $\alpha = 1$ . For category transitions,  $f(x)$  represents the proportion moving, not the proportion that have moved. This selectivity was designed for use in an age-based model to model maturity. In such a model, a logistic-producing maturation selectivity will (in the absence of other influences) make the proportions mature follow a logistic curve with parameters  $a_{50}$ ,  $a_{t095}$ .

**4.10.9. double\_normal**

$$f(x) = \begin{cases} \alpha 2^{-[(x-\mu)/\sigma_L]^2}, & \text{if } x \leq \mu \\ \alpha 2^{-[(x-\mu)/\sigma_R]^2}, & \text{if } x \geq \mu \end{cases} \quad (4.26)$$

The double-normal selectivity has estimable parameters  $a_1$ ,  $s_L$ , and  $s_R$ .  $\alpha$  is a scaling parameter, with default value of  $\alpha = 1$ . It has values  $\alpha$  at  $x = a_1$ , and  $0.5\alpha$  at  $x = a_1 - s_L$  and  $x = a_1 + s_R$ .

**4.10.10. double\_exponential**

$$f(x) = \begin{cases} \alpha y_0 (y_1/y_0)^{(x-x_0)/(x_1-x_0)}, & \text{if } x \leq x_0 \\ \alpha y_0 (y_2/y_0)^{(x-x_0)/(x_2-x_0)}, & \text{if } x > x_0 \end{cases} \quad (4.27)$$

The double-exponential selectivity has non-estimable parameters  $x_1$  and  $x_2$ , and estimable parameters  $x_0$ ,  $y_0$ ,  $y_1$ , and  $y_2$ .  $\alpha$  is a scaling parameter, with default value of  $\alpha = 1$ . It can be ‘U-shaped’. Bounds for  $x_0$  must be such that  $x_1 < x_0 < x_2$ . With  $\alpha = 1$ , the selectivity passes through the points  $(x_1, y_1)$ ,  $(x_0, y_0)$ , and  $(x_2, y_2)$ . If both  $y_1$  and  $y_2$  are greater than  $y_0$  the selectivity is ‘U-shaped’ with minimum at  $(x_0, y_0)$ .

**4.10.11. spline**

The spline selectivity implements a cubic spline that has non-estimable knots, and an estimable value for each knot. The cubic spline is either (i) a natural splines where the second derivatives are set to 0 at the boundaries, i.e., the values at the boundaries are horizontal, (ii) a spline with a fixed first derivative at the boundaries (linear, but not necessarily horizontal) and (iii) spline which turns into a parabola at the boundaries.

**4.11. Time Varying Parameters**

CASAL<sup>2</sup> has the functionality to vary a parameter between the start and final year of a model run. This can be for blocks of years or specific years if chosen. For years that are not specified the parameter will default to the input or if in a iterative state such as estimation mode, the value being trialled at that iteration. Available methods for time varying a parameter.

**4.11.1. Constant**

Allows a parameter to have an alternative values during the certain years, which is estimable.

**4.11.2. Random Walk**

A random deviate added into the last value drawn from a standard normal distribution. This has an estimable parameter  $\sigma$ . Caution although this has been crudely implemented it may not be a good idea to use as this is a random process, replicating it is not doable at current.

**4.11.3. Exogenous**

parameters are shifted based on an exogenous variable, an example of this is fishing selectivity parameters that may vary between years based on known changed fishing behaviours such as fishing season start time.

$$\delta_y = a(E_y - \bar{E}) \tag{4.28}$$

where  $\delta_y$  is the shift in parameter  $X$  in year  $y$ ,  $a$  is an estimable shift parameter,  $E$  is the exogenous variable and  $E_y$  is the value of this variable in year  $y$ . For more information readers can see Francis (2003)





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## 5. The estimation section

### 5.1. Role of the estimation section

The role of the estimation section is to define the tasks carried out by CASAL<sup>2</sup>:

1. Define the objective function (see Section 5.2)
2. Define the parameters to be estimated (see Section 5.3)
3. Calculate a point estimate, i.e., the maximum posterior density estimate (MPD) (see Section 5.4).
4. Calculate a posterior profile selected parameters, i.e., find, for each of a series of values of a parameter, allowing the other estimated parameters to vary, the minimum value of the objective function (see Section 5.5).
5. Generate an MCMC sample from the posterior distribution (see Section 5.6).
6. Calculate the approximate covariance matrix of the parameters as the inverse of the minimizer's approximation to the Hessian, and the corresponding correlation matrix (see Section 5.4).

The estimation section defines the objective function, parameters of the model, and the method of estimation (point estimates, Bayesian posteriors, profiles, etc.). The objective function is based on a goodness-of-fit measure of the model to observations, priors and penalties. See the observation section for a description of the observations, likelihoods, priors and penalties.

### 5.2. The objective function

*check and confirm text*

*KL comment:* This text from SPM. Does not cover all text in S6.7 of CASAL, e.g. max likelihood. OK to use? Additions required?

In Bayesian estimation, the objective function is a negative log-posterior,

$$Objective(p) = - \sum_i \log [L(\mathbf{p}|O_i)] - \log [\pi(\mathbf{p})] \quad (5.1)$$

where  $\pi$  is the joint prior density of the parameters  $p$ .

The contribution to the objective function from the likelihoods are defined in Section 6.1. In addition to likelihoods, priors (see Section 5.7) and penalties (see Section 5.8) are components of the objective function.

Penalties can be used to ensure that the exploitation rate constraints on mortality events (i.e., fisheries) are not breached (otherwise there is nothing to prevent the model from having abundances so low that the recorded mortalities could not have been taken), penalties on category transitions (to ensure there are enough individuals to move), and possibly penalties to encourage estimated values to be similar or smooth, etc.

### 5.3. Specifying the parameters to be estimated

The estimable parameters that will be estimated are defined using `@estimate` commands (see Section 9). An `@estimate` command-block looks like,

```
@estimate process[MyRecruitment].r0
lower_bound 1000
upper_bound 100000
type uniform
```

See Section 3.5.5 for instructions on how to generate the parameter name. At least one parameter to be estimated if doing an estimation, profile, or MCMC run. Initial values for the parameters to be estimated will still need to be provided, and these are used as the starting values for the minimiser. However, these may be overwritten if you provide a set of alternative starting values (i.e., using `casal2 -i`, see Section 3.4).

All parameters are estimated within bounds. For each parameter to be estimated, you need to specify the bounds and the prior (Section 5.7). Note that the bounds and prior for each parameter refer to the values of the parameters, not the actual values resulting from the application of the parameter to an equation. **Bounds should be carefully chosen as they effect the space in which the minimisers search over. Minimisers convert lower and upper bound to -1,1 space. Needs to be referenced.** If estimating only some elements of a vector, either define the elements of the vector to be estimated (see 3.5.5) or fix the others by setting the bounds equal.

The estimation of parameters can be phased. Here, some of the estimated parameters are initially held fixed, and a minimisation is carried out. Next, some or all of the remaining parameters that were initially held fixed are freed, and another minimisation is carried out. This process continues until all phases have been carried out.

#### 5.4. Point estimation

Point estimation is invoked with `casal2 -e`. Mathematically, it is an attempt to find a minimum of the objective function. CASAL<sup>2</sup> has multiple algorithms for solving (minimising) the optimisation problem. There are three non auto differential minimisers: numerical differences, differential evolution minimiser, and the dlib minimiser. There are also three auto differential minimisers being: ADOL-C, CPPAD, and BETADIFF. For references see section 1.7

##### 5.4.1. The numerical differences minimiser

The minimiser has three kinds of (non-error) exit status:

1. Successful convergence (suggests you have found a local minimum, at least).
2. Convergence failure (you have not reached a local minimum, though you may deem yourself to be 'close enough' at your own risk).
3. Convergence unclear (the minimiser halted but was unable to determine if convergence occurred. You may be at a local minimum, although you should check by restarting the minimiser at the final values of the estimated parameters).

You can choose the maximum number of quasi-Newton iterations and objective function evaluations allotted to the minimiser. If it exceeds either limit, it exits with a convergence failure. We recommend large numbers of evaluations and iterations (at least the defaults of 300 and 1000) unless you successfully reach convergence with less. You can also specify an alternative starting point of the minimiser using `casal2 -i`.

We want to stress that this is a local optimisation algorithm trying to solve a global optimisation problem. What this means is that, even if you get a 'successful convergence' message, your solution

may be only a local minimum, not a global one. To diagnose this problem, try doing multiple runs from different starting points and comparing the results, or doing profiles of one or more key parameters and seeing if any of the profiled estimates finds a better optimum than the original point estimate.

The approximate covariance matrix of the estimated parameters can be calculated as the inverse of the minimiser's approximation to the Hessian, and the corresponding correlation matrix is also calculated. Be aware that

- the Hessian approximation develops over many minimiser steps, so if the minimiser has only run for a small number of iterations the covariance matrix can be a very poor approximation
- the inverse Hessian is not a good approximation to the covariance matrix of the estimated parameters, and may not be useful to construct, for example, confidence intervals.

Also note that if an estimated parameter has equal lower and upper bounds, it will have entries of '0' in the covariance matrix and NaN or -1.#IND (depending on the operating system) in the correlation matrix.

#### 5.4.2. The differential evolution minimiser

The differential evolution minimiser is a simple population based, stochastic function minimizer, but is claimed to be quite powerful in solving minimisation problems. It is a method of mathematical optimization of multidimensional functions and belongs to the class of evolution strategy optimizers. Initially, the procedure randomly generates and evaluates a number of solution vectors (the population size), each with  $p$  parameters. Then, for each generation (iteration), the algorithm creates a candidate solution for each existing solution by random mutation and uniform crossover. The random mutation generates a new solution by multiplying the difference between two randomly selected solution vectors by some scale factor, then adding the result to a third vector. Then an element-wise crossover takes place with probability  $P_{cr}$ , to generate a potential candidate solution. If this is better than the initial solution vector, it replaces it, otherwise the original solution is retained. The algorithm is terminated after either a predefined number of generations (`max_generations`) or when the maximum difference between the scaled individual parameters from the candidate solutions from all populations is less than some predefined amount `tolerance`.

The differential evolution minimiser can be good at finding global minimums in surfaces that may have local minima. However, the speed of the minimiser, and the ability to find a good minima depend on the number of initial 'populations'. Some authors recommend that the number of populations be set at about  $10 * p$ , where  $p$  is the number of free parameters. However, depending on your problem, you may find that you may need more, or that less will suffice.

We note that there is no proof of convergence for the differential evolution solver, but several papers have found it to be an efficient method of solving multidimensional problems. Our (limited) experience suggests that it can often find a better minima and may be faster or longer (depending on the actual model specification) at finding a solution when compared with the numerical differences minimiser. Comparisons with auto-differentiation minimisers or other more sophisticated algorithms have not been made.

#### 5.4.3. Betadiff minimiser

An auto-differentiable minimiser for non-linear models.

#### 5.4.4. ADOL-C minimiser

An auto-differentiable minimiser for non-linear models.

#### 5.4.5. CPPAD minimiser

An auto-differentiable minimiser for non-linear models.

#### 5.4.6. Dlib minimiser

Non auto-diff minimiser

### 5.5. Posterior profiles

If profiles are requested `casal2 -p`, CASAL<sup>2</sup> will first calculate a point estimate. For each scalar parameter or, in the case of vectors or selectivities, the element of the parameter to be profiled, CASAL<sup>2</sup> will fix its value at a sequence of  $n$  evenly spaced numbers (*step*) between a specified lower and upper bounds  $l$  and  $u$ , and calculate a point estimate at each value.

By default  $step = 10$ , and  $(l, u) = (\text{lower bound on parameter plus } (range/(2n)), \text{upper bound on parameter less } (range/(2n)))$ . Each minimisation starts at the final parameter values from the previous resulting value of the parameter being profiled. CASAL<sup>2</sup> will report the objective function for each parameter value. Note that an initial point estimate should be compared with the profile, not least to check that none of the other points along the profile have a better objective function value than the initial ‘minimum’.

You specify which parameters are to be profiled, and optionally the number of steps, lower bound, and upper bound for each. In the case of vector parameters, you will also need to specify the element of the vector being profiled.

You can also supply the initial starting point for the estimation using `casal2 -i file` — this may improve the minimiser performance for the profiles.

If you get an implausible profile, it may be a result of not using enough iterations in the minimiser or a poor choice of minimiser control variables (e.g., the minimiser tolerance). It also may be useful to try both if the minimisers in CASAL<sup>2</sup> and compare the results.

### 5.6. Bayesian estimation

*check and confirm text*

*KL comment: This text is from SPM and is nearly verbatim S6.5 in CASASL, but two large sections excluded: ...request covariate matrix change adaptively... and from ...multivariate t dist... onwards. OK to use/ Additions required?*

CASAL<sup>2</sup> can use a Monte Carlo Markov Chain (MCMC) to generate a sample from the posterior distribution of the estimated parameters `casal2 -m` and output the sampled values to a file (optionally keeping only every  $n$ th set of values).

As CASAL<sup>2</sup> has no post-processing capabilities. CASAL<sup>2</sup> cannot produce MCMC convergence diagnostics (use a package such as BOA) or plot/summarize the posterior distributions of the output quantities (for example, using a general-purpose statistical or spreadsheet package such as S-Plus, R, or Microsoft Excel).

Bayesian methodology and MCMC are both large and complex topics, and we do not describe either properly here. See Gelman et al. (1995) and Gilks et al. (1994) for details of both Bayesian analysis and MCMC methods. In addition, see Punt & Hilborn (2001) for an introduction to quantitative fish stock assessment using Bayesian methods.

This section only briefly describes the MCMC algorithms used in CASAL<sup>2</sup>. See Section 9.3 for a better description of the sequence of CASAL<sup>2</sup> commands used in a full Bayesian analysis.

CASAL<sup>2</sup> uses a straightforward implementation of the Metropolis-Hastings algorithm (Gelman et al., 1995, Gilks et al., 1994). The Metropolis-Hastings algorithm attempts to draw a sample from a Bayesian posterior distribution, and calculates the posterior density  $\pi$ , scaled by an unknown constant. The algorithm generates a ‘chain’ or sequence of values. Typically the beginning of the chain is discarded and every  $N$ th element of the remainder is taken as the posterior sample. The chain is produced by taking an initial point  $x_0$  and repeatedly applying the following rule, where  $x_i$  is the current point:

- Draw a candidate step  $s$  from a proposal distribution  $J$ , which should be symmetric i.e.,  $J(-s) = J(s)$ .
- Calculate  $r = \min(\pi(x_i + s)/\pi(x_i), 1)$ .
- Let  $x_{i+1} = x_i + s$  with probability  $r$ , or  $x_i$  with probability  $1 - r$ .

An initial point estimate is produced before the chain starts, which is done so as to calculate the approximate covariance matrix of the estimated parameters (as the inverse Hessian), and may also be used as the starting point of the chain.

The user can specify the starting point of the point estimate minimiser using `casal2 -i`. Don’t start it too close to the actual estimate (either by using `casal2 -i`, or by changing the initial parameter values in input configuration file) as it takes a few iterations to form a reasonable approximation to the Hessian.

There are two options for the starting point of the Markov Chain:

- Start from the point estimate.
- Start from a random point near the point estimate (the point is generated from a multivariate normal distribution, centred on the point estimate, with covariance equal to the inverse Hessian times a user-specified constant). This may be useful if the chain gets ‘stuck’ at the point estimate, or if you wish to generate multiple chains from for later MCMC diagnostic tests.
- Start from a point specified by the user with `casal2 -i` (*was NYI, to be included?*)

The chain moves in natural space, i.e., no transformations are applied to the estimated parameters. The default proposal distribution is a multivariate  $t$  centred on the current point, with covariance matrix equal to a matrix based on the approximate covariance produced by the minimiser, times some stepsize factor. The following steps define the initial covariance matrix of the proposal distribution:

- The covariance matrix is taken as the inverse of the approximate Hessian from the quasi-Newton minimiser.
- The covariance matrix is modified so as to decrease all correlations greater than `@mcmc.max_correlation` down to `@mcmc.max_correlation`, and similarly to increase all correlations less than `-@mcmc.max_correlation` up to `-@mcmc.max_correlation` (the `@mcmc.max_correlation` parameter defaults to 0.8). This should help to avoid getting ‘stuck’ in a lower-dimensional subspace.

- The covariance matrix is then modified either by,
  - if `@mcmc.adjustment_method=covariance`: that if the variance of the  $i$ th parameter is non-zero and less than `@mcmc.min_difference` times the difference between the parameters' lower and upper bound, then the variance is changed, without changing the associated correlations, to  $k = \min\_diff(upper\_bound_i - lower\_bound_i)$ . This is done by setting

$$\text{Cov}(i, j)' = \text{sqrt}(k) \text{Cov}(i, j) / \text{sd}(i)$$

for  $i \neq j$ , and  $\text{var}(i)' = k$

- if `@mcmc.adjustment_method=correlation`: that if the variance of the  $i$ th parameter is non-zero and less than `@mcmc.min_difference` times the difference between the parameters' lower and upper bound, then its variance is changed to  $k = \min\_diff(upper\_bound_i - lower\_bound_i)$ . This differs from (i) above in that the effect of this option is that it also modifies the resulting correlations between the  $i$ th parameter and all other parameters.

This allows each estimated parameter to move in the MCMC even if its variance is very small according to the inverse Hessian. In both cases, the `@mcmc.min_difference` parameter defaults to 0.0001.

- The `@mcmc.stepsize` (a scalar factor applied to the covariance matrix to improve the acceptance probability) is chosen by the user. The default is  $2.4d^{-0.5}$  where  $d$  is the number of estimated parameters, as recommended by Gelman et al. (Gelman et al., 1995). However, you may find that a smaller value may often be better.

The proposal distribution can also change adaptively during the chain, using two different mechanisms. Both are offered as means of improving the convergence properties of the chain. It is important to note that any adaptive behaviour must finish before the end of the burn-in period, i.e., the proposal distribution must be finalised before the kept portion of the chain starts. The adaptive mechanisms are as follows:

1. You can request that the stepsize change adaptively at one or more sample numbers. At each adaptation, the stepsize is doubled if the acceptance rate since the last adaptation is more than 0.5, or halved if the acceptance rate is less than 0.2. (See Gelman et al. (Gelman et al., 1995) for justification.)
2. You can request that the entire covariance matrix change adaptively at one or more sample numbers. At each adaptation, it is replaced with a matrix based on the sample covariance of an earlier section of the chain. The theory here is that the covariance of a portion of chain could potentially be a better estimate of the covariance of the posterior distribution than the inverse Hessian. *(was NYI, to be included?)*

The procedure used to choose the sample of points is as follows. First, all points on the chain so far are taken. All points in an initial user-specified period are discarded. The assumption is that the chain will have started moving during this period - if this is incorrect and the chain has still not moved by the end of this period, it is a fatal error and CASAL<sup>2</sup> stops. The remaining set of points must contain at least some user-specified number of transitions - if this is incorrect and the chain has not moved this often, it is again a fatal error. If this test is passed, the set of points is systematically sub-sampled down to 1000 points (it must be at least this long to start with). *(was NYI, to be included?)*

The variance-covariance matrix of this sub-sample of chain is calculated. As above, correlations greater than `@mcmc.max_correlation` are reduced to `@mcmc.max_correlation`, correlations less than `@mcmc.max_correlation` are increased to `@mcmc.max_correlation`, and very small non-zero variances are increased (`@mcmc.covariance_adjustment` and `@mcmc.min_difference`). The result is the new variance-covariance matrix of the proposal distribution. *(was NYI, to be included?)*

The stepsize parameter is now on a completely different scale, and must be reset. It is set to a user-specified value (which may or may not be the same as the initial stepsize). We recommend that some of the stepsize adaptations are set to occur after this, so that the stepsize can be readjusted to an appropriate value which gives good acceptance probabilities with the new matrix. *(was NYI, to be included?)*

All modified versions of the covariance matrix are printed to the standard output, but only the initial covariance matrix (inverse Hessian) is saved to the objectives file. The number of covariance modifications by each iteration is recorded as a column on the objectives file. *(was NYI, to be included?)*

The probability of acceptance for each jump is 0 if it would move out of the bounds, or 1 if it improves the posterior, or (new posterior/old posterior) otherwise. You can specify how often the position of the chain is recorded using the keep parameter. For example, with keep 10, only every 10th sample is recorded.

You have the option to specify that some of the estimated parameters are fixed during the MCMC. If the chain starts at the point estimate or at a random location, these fixed parameters are set to their values at the point estimate.

If you specify the start of the chain using `casal2 -i`, these fixed parameters are set to the values in the file. *(was NYI, to be included?)*

The posterior sample can be used for (projections (Section 4.4.3)*(was NYI, to be included?)*) or simulations (Section 6.7) with the values supplied using `casal2 -i file`.

*(following from CASAL, to be included?)*

A multivariate t distribution is available as an alternative to the multivariate normal proposal distribution. If you request multivariate t proposals, you may want to change the degrees of freedom from the default of 4. As the degrees of freedom decrease, the t distribution becomes more heavy tailed. This may lead to better convergence properties.

Having produced one or more Markov chains and looked at the diagnostics, reload all the chain output files into CASAL and use them to generate a single posterior sample (using `-C`). At this stage, the first `burn_in` iterations for each chain are discarded (so, with keep 10, `burn_in` 1000, the first 1000 recorded samples are discarded for each chain). Unless a very large value of keep was originally chosen, it will be necessary to further reduce the size of the posterior sample (possibly down to several hundred) such that it can be analysed in a reasonable amount of time. This is done by sub-sampling. You specify the size of the sub-sample to be produced (or else no sub-sampling is done). You have the option to generate a systematic sub-sample (i.e., every `nth` point is kept) or a random sub-sample (the former is recommended except with prior re-weighting, when the latter must be used).

Given a posterior (sub)sample, CASAL can calculate a list of output quantities for each sample point (see Section 7.2). These quantities can be dumped into a file (using `casal -v`) and read into an external software package where the posterior distributions can be plotted and/or summarised.

The posterior sample can also be used for projections (Section 7.3) and stochastic yield calculations (Section 7.5). The advantage of this is that the parameter uncertainty, as expressed in your posterior distribution, can be included into the risk and yield estimates.

It is possible to investigate the results that would have been obtained if a different prior had been specified. This is called prior re-weighting and is done by calculating the ratio of the new prior to the original prior for each point in the posterior sample, then using these ratios as probability weights when generating a random (not systematic) sub-sample with `casal -C`. Prior re-weighting is applicable only if the new prior is zero in every part of the parameter space for which the original prior was zero. Also, it is likely to be numerically unstable unless the new prior is very small in every part of the parameter space for which the original prior was very small.

## 5.7. Priors

In a Bayesian analysis, you need to give a prior for every parameter that is being estimated. There are no default priors.

Note that when some of these priors are parameterised in terms of mean, c.v., and standard deviation, these refer to the parameters of the distribution before bounds are applied. The moments of the prior after the bounds are applied may differ.

CASAL<sup>2</sup> has the following priors (expressed in terms of their contribution to the objective function):

1. Uniform

$$-\log(\pi(p)) = 0 \tag{5.2}$$

2. Uniform-log (i.e.,  $\log(p) \sim \text{uniform}$ )

$$-\log(\pi(p)) = \log(p) \tag{5.3}$$

3. Normal with mean  $\mu$  and c.v.  $c$

$$-\log(\pi(p)) = 0.5 \left( \frac{p - \mu}{c\mu} \right)^2 \tag{5.4}$$

4. Normal with mean  $\mu$  and standard deviation  $\sigma$

$$-\log(\pi(p)) = 0.5 \left( \frac{p - \mu}{\sigma} \right)^2 \tag{5.5}$$

5. Lognormal with mean  $\mu$  and c.v.  $c$

$$-\log(\pi(p)) = \log(p) + 0.5 \left( \frac{\log(p/\mu)}{s} + \frac{s}{2} \right)^2 \tag{5.6}$$

where  $s$  is the standard deviation of  $\log(p)$  and  $s = \sqrt{\log(1 + c^2)}$ .

(following from CASAL, to be included?)



6. Normal-log with  $\log(p)$  having mean  $m$  and standard deviation  $s$ ,

6. Beta with mean  $\mu$  and standard deviation  $\sigma$ , and range parameters  $A$  and  $B$

$$-\log(\pi(p)) = (1-m)\log(p-A) + (1-n)\log(B-p) \quad (5.7)$$

where  $v = \frac{\mu-A}{B-A}$ , and  $\tau = \frac{(\mu-A)(B-\mu)}{\sigma^2} - 1$  and then  $\mu = \tau v$  and  $n = \tau(1-v)$ . Note that the beta prior is undefined when  $\tau \leq 0$ .

*(following from CASAL, to be included?)*

Vectors of parameters can be independently (but not necessarily identically) distributed according to any of the above forms, in which case the joint negative-log-prior for the vector is the sum of the negative-log-priors of the components. Values of each parameter need to be specified for each element of the vector.

In addition, for a vector  $p$  of  $n$  identically distributed parameters (for example, YCS) the following priors are allowed:

1. Multivariate normal from a stationary AR(1) process with parameters
  - .
  - .
  - .
2. Multivariate normal-log, where  $\log(p)$  forms a stationary AR(1) process as per 1. above, with parameters
  - .
  - .
  - .
3. Multivariate normal-log with mean 1, where  $E(p_i)=1$  and  $\log(p)$  forms a stationary AR(1) process as for the multivariate normal above, with parameters
  - .
  - .
  - .

## 5.8. Penalties

Penalties are associated with processes and can be used to encourage or discourage parameter values or model outputs that are unlikely to be sensible, by adding a penalty to the objective function. For example, parameter estimates that do not allow a known mortality event to remove enough individuals from the population can be discouraged with an event mortality penalty. CASAL<sup>2</sup> requires penalty functions for processes that move or shift a *number* of individuals between categories or from the partition.

For most penalties, you need to specify a multiplier, and the objective function is increased by this multiplier times the penalty value as described below. In some cases you will need to make the multiplier quite large to prohibit some model behaviour.

Currently, the penalties for the processes `@process[label].type=event_mortality`, `@process[label].type=tag_by_length` and `@process[label].type=category_transition` are the only penalties implemented.

For these processes, two types of penalty can be defined, natural scale (the default) and log scale. Both of these types add a penalty value of the squared difference between the observed value (i.e., the actual number of individuals to be removed in an event mortality process or the actual number

of individuals to shift in a category transition process), and the number that were moved (if less than or equal), times the penalty multiplier.

The natural scale penalty just uses at the squared difference on a natural scale, while the log scale penalty uses the squared difference of the logged values.

### 5.9. Additional Priors

Additional priors are the inverse

### 5.10. Estimate Transformations

CASAL<sup>2</sup> has the untested functionality of transforming an estimated parameter in a new space. This may be done to remove correlation for other convergence or optimisation purposes. This functionality transforms the estimate and the bounds to the transformed space along with the prior. To account for the change variable a Jacobian is added to the objective function. For more information uses are asked to read the STAN manual **REFERENCE**. The user must supply the type, bounds for the transformed variable can be supplied by the user, but if not CASAL<sup>2</sup> will work them out. NOTE must be used with caution. May be buggy!!!

#### 5.10.1. log

#### 5.10.2. Inverse

#### 5.10.3. Log odds

#### 5.10.4. Simplex

---

## 6. The observation section

### 6.1. Observations and likelihoods

Observations are typically supplied as observations at an instance in time, over some spatially aggregated area. Time series of observations can be supplied as separate observations for each year or point in time.

CASAL<sup>2</sup> allows the following types of observations;

- Observations of proportions by age class within categories
- Observations of proportions between categories within age classes
- Relative and absolute abundance/biomass observations

The definitions for each type of observation are described below, including how the observed values should be supplied, how CASAL<sup>2</sup> calculates the expected values, and the likelihoods that are available for each type of observation.

CASAL<sup>2</sup> evaluates the observations at the end of a time-step (i.e., after all of the processes for that time-step have been applied). However, the observation can be applied to the abundance at the start of a time-step or part-way through a time-step by the use of the `proportion_time_step` subcommand.

By default (i.e., if `proportion_method = mean`), the partition at some point  $p$  during the time-step is then evaluated as the weighted sum between the start and end of the time-step, i.e., for any element  $i$  in the partition,  $n_i = (1 - p)n_i^{start} + pn_i^{end}$ . Note that it may not be sensible to use a value other than one, depending on the processes that happen during the time-step (for example, if the time-step contains an ageing process).

If the `proportion_method = difference`, then the observation is of the *difference* between the population state at the start of the time-step and the end. This can be used to generate expected values for observations of, for example removals due to a mortality event, by only having a single process in the time-step. In this case, the `proportion_time_step` is simply a multiplier of the population state.

### 6.2. Proportions-at-age observations

Proportions-at-age observations are observations of either the relative number of individuals at age or relative biomass at age, via some selectivity.

The observation is supplied for a given year and time-step, for some selected age classes of the population (i.e., for a range of ages multiplied by a selectivity), for categories aggregated over a set of spatial cells. Note that the categories defined in the observations must have an associated selectivity, defined by `selectivities`.

The age range must be ages defined in the partition (i.e., between `@model.min_age` and `@model.max_age` inclusive), but the upper end of the age range can optionally be a plus group — which must be either the same or less than the plus group defined for the partition.

Proportions-at-age observations can be supplied as;

1. a set of proportions for a single category,
2. a set of proportions for multiple categories, or
3. a set of proportions across aggregated categories.

For example, for a model with the two categories *male* and *female*, we might supply either (i) a set of proportions for a single category (i.e., males) within each age class; (ii) a set of proportions describing the proportions of individuals within each age class across multiple categories (i.e., males and females) simultaneously, or (iii) a set of proportions for the total number of individuals over the aggregated categories (i.e., males + females) combined, within each age class.

The way the categories of the observation are defined specifies which of these alternatives are used. It is also possible to have an observation with multiple and aggregated categories simultaneously.

### Proportions-at-age for a single category

This form of defining the observation is the simplest, and is used to model a set of proportions of a single category by age class. For example, to specify that the observations are of the proportions of male within each age class, then the subcommand `categories` for the `@observation[label].type=proportion.by_age` command is,

```
categories male
```

CASAL<sup>2</sup> then expects that there will be a single vector of proportions supplied, with one proportion for each age class within the defined age range, and that these proportions sum to one.

For example, if the age range was 3 to 10, then 8 proportions should be supplied (one proportion for each of the ages 3, 4, 5, 6, 7, 8, 9, and 10). The expected values will be the expected proportions of males within each of these age classes (after ignoring any males aged less than 3 or older than 10), after applying a selectivity at the year and time-step specified. The supplied vector of proportions (i.e., in this example, the 8 proportions) must sum to one, which is evaluated with a default tolerance of 0.001.

The observations must be also supplied using all or some of the values of defined by some *categorical* layer. CASAL<sup>2</sup> calculates the expected values by summing over the defined ages (via the age range and selectivity) and categories for those spatial cells where the categorical layer has the same value as defined for each vector of observations.

For example, in a  $2 \times 2$  spatial model a categorical layer (e.g., with label *Area*) may define that cells (1,1) and (1,2) have value *A* and cells (2,1) and (2,2) have value *B*, i.e.,

```
@layer Area
type categorical
data A A
data B B
```

The observations for those spatial cells where the categorical layer has value *A* would be,

```
@observation MyProportions
type proportions_at_age
layer Area
...
categories male
min_age 1
max_age 5
obs A 0.01 0.09 0.20 0.30 0.40
...
```

Or, for both *A* and *B* as,

```
@observation MyProportions
type proportions_at_age
layer Area
...
categories male
min_age 1
max_age 5
obs A 0.01 0.09 0.20 0.30 0.40
obs B 0.02 0.06 0.12 0.25 0.55
...
```

Note that to have an observation for each individual spatial cell in a model, then define a categorical layer that has a single, unique value for each spatial cell for use in the observation.

### Proportions-at-age for multiple categories

This form of the observation extends the idea above for multiple categories. It is used to model a set of proportions over several categories by age class. For example, to specify that the observations are of the proportions of male or females within each age class, then the subcommand `categories` for the `@observation[label].type=proportion_by_age` command is,

```
categories male female
```

CASAL<sup>2</sup> then expects that there will be a single vector of proportions supplied, with one proportion for each category and age class combination, and that these proportions sum to one.

For example, if there were two categories and the age range was 3 to 10, then 16 proportions should be supplied (one proportion for each of the the ages 3, 4, 5, 6, 7, 8, 9, and 10, for each category male and female). The expected values will be the expected proportions of males and within each of these age classes (after ignoring those aged less than 3 or older than 10), after applying a selectivity at the year and time-step specified. The supplied vector of proportions (i.e., in this example, the 16 proportions) must sum to one, which is evaluated with a default tolerance of 0.001.

For example, using the earlier spatial model with a categorical layer that has label *Area*, the observations for those spatial cells where the categorical layer has value *A* would be,

```
@observation MyProportions
type proportions_at_age
layer Area
...
categories male female
min_age 1
max_age 5
obs A 0.01 0.05 0.10 0.20 0.20 0.01 0.05 0.15 0.20 0.03
obs B 0.02 0.06 0.10 0.21 0.18 0.02 0.05 0.15 0.20 0.01
...
```

### Proportions-at-age across aggregated categories

This form of the observation extends the idea above, but allows categories to be aggregated before the proportions are calculated. It is used to model a set of proportions from several categories that have been combined by age class. To indicate that two (or more) categories are to be aggregated,

separate them with a '+' symbol. For example, to specify that the observations are of the proportions of male and females combined within each age class, then the subcommand `categories` for the `@observation[label].type=proportion_by_age` command is,

```
categories male + female
```

CASAL<sup>2</sup> then expects that there will be a single vector of proportions supplied, with one proportion for each age class, and that these proportions sum to one.

For example, if there were two categories and the age range was 3 to 10, then 8 proportions should be supplied (one proportion for each of the ages 3, 4, 5, 6, 7, 8, 9, and 10, for the sum of males and females within each age class). The expected values will be the expected proportions of males + females within each of these age classes (after ignoring those aged less than 3 or older than 10), after applying a selectivity at the year and time-step specified. The supplied vector of proportions (i.e., in this example, the 16 proportions) must sum to one, which is evaluated with a default tolerance of 0.001.

For example, using the earlier spatial model with a categorical layer that has label `Area`, the observations for those spatial cells where the categorical layer has value `A` would be,

```
@observation MyProportions
type proportions_at_age
layer Area
...
categories male + female
min_age 1
max_age 5
obs A 0.02 0.13 0.25 0.30 0.30
obs B 0.02 0.06 0.18 0.35 0.39
...
```

The later form can then be extended to include multiple categories, or multiple aggregated categories. For example, to describe proportions for the three groups: immature males, mature males, and all females (immature and mature females added together) for ages 1–4, a total of 12 proportions are required

```
@observation MyProportions
type proportions_at_age
layer Area
...
categories male_immature male_mature female_immature + female_mature
min_age 1
max_age 4
obs A 0.05 0.15 0.15 0.05 0.02 0.03 0.08 0.04 0.05 0.15 0.15 0.08
...
```

### 6.2.1. Likelihoods for proportions-at-age observations

CASAL<sup>2</sup> implements two likelihoods for proportions-at-age observations, the multinomial likelihood and the lognormal likelihood.

#### The multinomial likelihood

For the observed proportions at age  $O_i$  for age classes  $i$ , with sample size  $N$ , and the expected proportions at the same age classes  $E_i$ , the negative log-likelihood is defined as;

$$-\log(L) = -\log(N!) + \sum_i \log((NO_i)!) - NO_i \log(Z(E_i, \delta)) \quad (6.1)$$

where  $\sum_i O_i = 1$  and  $\sum_i E_i = 1$ .  $Z(\theta, \delta)$  is a robustifying function to prevent division by zero errors, with parameter  $\delta > 0$ .  $Z(\theta, \delta)$  is defined as,

$$Z(\theta, \delta) = \begin{cases} \theta, & \text{where } \theta \geq r \\ \delta / (2 - \theta / \delta), & \text{otherwise} \end{cases} \quad (6.2)$$

The default value of  $\delta$  is  $1 \times 10^{-11}$ .

### The lognormal likelihood

For the observed proportions at age  $O_i$  for age classes  $i$ , with c.v.  $c_i$ , and the expected proportions at the same age classes  $E_i$ , the negative log-likelihood is defined as;

$$-\log(L) = \sum_i \left( \log(\sigma_i) + 0.5 \left( \frac{\log(O_i / Z(E_i, \delta))}{\sigma_i} + 0.5 \sigma_i \right)^2 \right) \quad (6.3)$$

where

$$\sigma_i = \sqrt{\log(1 + c_i^2)} \quad (6.4)$$

and the  $c_i$ 's are the c.v.s for each age class  $i$ , and  $Z(\theta, \delta)$  is a robustifying function to prevent division by zero errors, with parameter  $\delta > 0$ .  $Z(\theta, \delta)$  is defined as,

$$Z(\theta, \delta) = \begin{cases} \theta, & \text{where } \theta \geq r \\ \delta / (2 - \theta / \delta), & \text{otherwise} \end{cases} \quad (6.5)$$

The default value of  $\delta$  is  $1 \times 10^{-11}$ .

## 6.3. Proportions-by-category observations

Proportions-by-category observations are observations of either the relative number of individuals between categories within age classes, or relative biomass between categories within age classes.

The observation is supplied for a given year and time-step, for some selected age classes of the population (i.e., for a range of ages multiplied by a selectivity), for categories aggregated over a set of spatial cells.

The age range must be ages defined in the partition (i.e., between `@model.min_age` and `@model.max_age` inclusive), but the upper end of the age range can optionally be a plus group — which may or may not be the same as the plus group defined for the partition.

Proportions-by-category observations can be supplied for any set of categories as a proportion of themselves and any set of additional categories. For example, for a model with the two categories *male* and *female*, we might supply observations of the proportions of males in the population at each age class. The subcommand `categories` defines the categories for the numerator in the calculation of the proportion, and the subcommand `categories2` supplies the additional categories to be used in the denominator of the calculation. In addition, each category must have an associated selectivity, defined by `selectivities` for the numerator categories and `selectivities2` for the additional categories used in the denominator, e.g.,

```
categories male
```

```
categories2 female
selectivities male-selectivity
selectivities2 female-selectivity
```

defines that the proportion of males in each age class as a proportion of males + females. CASAL<sup>2</sup> then expects that there will be a vector of proportions supplied, with one proportion for each age class within the defined age range, i.e., if the age range was 3 to 10, then 8 proportions should be supplied (one proportion for each of the ages 3, 4, 5, 6, 7, 8, 9, and 10). The expected values will be the expected proportions of male to male + female within each of these age classes, after applying the selectivities at the year and time-step specified.

The observations must be supplied using all or some of the values defined by a categorical layer. CASAL<sup>2</sup> calculates the expected values by summing over the ages (via the age range and selectivity) and categories for those spatial cells where the categorical layer has the same value as defined for each vector of observations.

For example, in a  $2 \times 2$  spatial model a categorical layer (e.g., with label Area) may define that cells (1,1) and (1,2) have value *A* and cells (2,1) and (2,2) have value *B*, i.e.,

```
@layer Area
type categorical
data A A
data B B
```

Here we supply observations for those spatial cells where the categorical layer has value *A* as,

```
@observation MyProportions
type proportions_by_category
layer Area
...
categories male
categories2 female
min_age 1
max_age 5
obs A 0.01 0.05 0.10 0.20 0.20
...
```

Or, for both *A* and *B* as,

```
@observation MyProportions
type proportions_by_category
layer Area
...
categories male
categories2 female
min_age 1
max_age 5
obs A 0.01 0.05 0.10 0.20 0.20
obs B 0.02 0.06 0.10 0.21 0.18
...
```

To supply an observation for individual spatial cells, then you will need to define a categorical layer with a single, unique value for each spatial cell.

### 6.3.1. Likelihoods for proportions-by-category observations

CASAL<sup>2</sup> implements two likelihoods for proportions-by-category observations, the binomial likelihood, and the normal approximation to the binomial (binomial-approx).



### The binomial likelihood

For observed proportions  $O_i$  for age class  $i$ , where  $E_i$  are the expected proportions for age class  $i$ , and  $N_i$  is the effective sample size for age class  $i$ , then the negative log-likelihood is defined as;

$$-\log(L) = -\sum_i [\log(N_i!) - \log((N_i(1 - O_i))!) - \log((N_i O_i)!) + N_i O_i \log(Z(E_i, \delta)) + N_i(1 - O_i) \log(Z(1 - E_i, \delta))] \quad (6.6)$$

where  $Z(\theta, \delta)$  is a robustifying function to prevent division by zero errors, with parameter  $\delta > 0$ .  $Z(\theta, \delta)$  is defined as,

$$Z(\theta, \delta) = \begin{cases} \theta, & \text{where } \theta \geq r \\ \delta / (2 - \theta / \delta), & \text{otherwise} \end{cases} \quad (6.7)$$

The default value of  $\delta$  is  $1 \times 10^{-11}$ .

### The normal approximation to the binomial likelihood

For observed proportions  $O_i$  for age class  $i$ , where  $E_i$  are the expected proportions for age class  $i$ , and  $N_i$  is the effective sample size for age class  $i$ , then the negative log-likelihood is defined as;

$$-\log(L) = \sum_i \log \left( \sqrt{Z(E_i, \delta) Z(1 - E_i, \delta) / N_i} \right) + \frac{1}{2} \left( \frac{O_i - E_i}{\sqrt{Z(E_i, \delta) Z(1 - E_i, \delta) / N_i}} \right)^2 \quad (6.8)$$

where  $Z(\theta, \delta)$  is a robustifying function to prevent division by zero errors, with parameter  $\delta > 0$ .  $Z(\theta, \delta)$  is defined as,

$$Z(\theta, \delta) = \begin{cases} \theta, & \text{where } \theta \geq r \\ \delta / (2 - \theta / \delta), & \text{otherwise} \end{cases} \quad (6.9)$$

The default value of  $\delta$  is  $1 \times 10^{-11}$ .

## 6.4. Abundance or biomass observations

Abundance (or biomass) observations are observations of either a relative or absolute number (or biomass) of individuals from a set of categories after applying a selectivity. The observations classes are the same, except that a biomass observation will use the biomass as the observed (and expected) value (calculated from mean weight of individuals within each age and category) while an abundance observation is just the number of individuals.

Each observation is for a given year and time-step, for some selected age classes of the population (i.e., for a range of ages multiplied by a selectivity), for categories aggregated over a set of spatial cells. Further, you need to provide the label of the catchability coefficient  $q$ , which can either be estimated or fixed. For absolute abundance or absolute biomass observations, define a catchability where  $q = 1$ .

The observations can be supplied for any set of categories. For example, for a model with the two categories *male* and *female*, we might supply an observation of the total abundance/biomass (male + female) or just male abundance/biomass. The subcommand `categories` defines the categories used to aggregate the abundance/biomass. In addition, each category must have an associated selectivity, defined by `selectivities`. For example,

```
categories male
selectivities male-selectivity
```

defines an observation for males after applying the selectivity male-selectivity. CASAL<sup>2</sup> then expects that there will be a single observation supplied. The expected values for the observations will be the expected abundance (or biomass) of males, after applying the selectivities, at the year and time-step specified.

The observations must be supplied using all or some of the values of defined by a categorical layer. CASAL<sup>2</sup> calculates the expected values by summing over the defined ages (via the age range and selectivity) and categories for those spatial cells where the categorical layer has the same value as defined for each vector of observations.

For example, in a  $2 \times 2$  spatial model a categorical layer (e.g., with label Area) may define that cells (1,1) and (1,2) have value *A* and cells (2,1) and (2,2) have value *B*, i.e.,

```
@layer Area
type categorical
data A A
data B B
```

Here we supply abundance observations for those spatial cells where the categorical layer has value *A* as,

```
@observation MyAbundance
type abundance
layer Area
...
categories male
obs A 1000
...
```

Or, for both *A* and *B* as,

```
@observation MyAbundance
type abundance
layer Area
...
categories male
obs A 1000
obs B 1200
...
```

To supply an observation for individual spatial cells, then you will need to define a categorical layer with a single, unique value for each spatial cell.

Note that, to define a biomass observation instead of an abundance observation, use

```
@observation MyBiomass
type biomass
...
```

### 6.4.1. Likelihoods for abundance observations

#### The lognormal likelihood

For observations  $O_i$ , c.v.  $c_i$ , and expected values  $qE_i$ , the negative log-likelihood is defined as;

$$-\log(L) = \sum_i \left( \log(\sigma_i) + 0.5 \left( \frac{\log(O_i/qZ(E_i, \delta))}{\sigma_i} + 0.5\sigma_i \right)^2 \right) \quad (6.10)$$

where

$$\sigma_i = \sqrt{\log(1 + c_i^2)} \quad (6.11)$$

and  $Z(\theta, \delta)$  is a robustifying function to prevent division by zero errors, with parameter  $\delta > 0$ .  $Z(\theta, \delta)$  is defined as,

$$Z(\theta, \delta) = \begin{cases} \theta, & \text{where } \theta \geq r \\ \delta / (2 - \theta / \delta), & \text{otherwise} \end{cases} \quad (6.12)$$

The default value of  $\delta$  is  $1 \times 10^{-11}$ .

#### The normal likelihood

For observations  $O_i$ , c.v.  $c_i$ , and expected values  $qE_i$ , the negative log-likelihood is defined as;

$$-\log(L) = \sum_i \left( \log(c_i E_i) + 0.5 \left( \frac{O_i - E_i}{Z(c_i E_i, \delta)} \right)^2 \right) \quad (6.13)$$

and  $Z(\theta, \delta)$  is a robustifying function to prevent division by zero errors, with parameter  $\delta > 0$ .  $Z(\theta, \delta)$  is defined as,

$$Z(\theta, \delta) = \begin{cases} \theta, & \text{where } \theta \geq r \\ \delta / (2 - \theta / \delta), & \text{otherwise} \end{cases} \quad (6.14)$$

The default value of  $\delta$  is  $1 \times 10^{-11}$ .

### 6.5. Process error

Additional ‘process error’ can be defined for each set of observations. Additional process error has the effect of increasing the observation error in the data, and hence of decreasing the relative weight given to the data in the fitting process.

For observations where the likelihood is parameterised by the c.v., you can specify the process error for a given set of observations as a c.v., in which case all the c.v.s  $c_i$  are changed to

$$c'_i = \sqrt{c_i^2 + c_{process\_error}^2} \quad (6.15)$$

Note that  $c_{process\_error} \geq 0$ , and that  $c_{process\_error} = 0$  is equivalent to no process error.

Similarly, if the likelihood is parameterised by the effective sample size  $N$ ,

$$N'_i = \frac{1}{1/N_i + 1/N_{process\_error}} \quad (6.16)$$

Note that this requires that  $N_{process\_error} > 0$ , but we allow the special case of  $N_{process\_error} = 0$ , and define  $N_{process\_error} = 0$  as no process error (i.e., defined to be equivalent to  $N_{process\_error} = \infty$ ).

For both the c.v. and  $N$  process errors, the process error has more effect on small errors than on large ones. Be clear that a large value for the  $N$  process error means a small process error.

## 6.6. Ageing error

CASAL<sup>2</sup> can apply ageing error age frequency observations. Ageing error is applied to the expected values for proportions-at-age observations. The ageing error is applied as a misclassification matrix, which has the effect of 'smearing' the age frequencies. These are used in calculating the fits to the observed values, and hence the contribution to the total objective function.

Ageing error is optional, and if it is used, it may be omitted for any individual time series. Different ageing error models may be applied for different observation commands. See Section 7.11 for reporting the misclassification matrix.

The ageing error models implemented are,

1. None: The default model is to apply no ageing error.
2. Off by one: Proportion  $p_1$  of individuals of each age  $a$  are misclassified as age  $a - 1$  and proportion  $p_2$  are misclassified as age  $a + 1$ . Individuals of age  $a < k$  are not misclassified. If there is no plus group in the population model, then proportion  $p_2$  of the oldest age class will 'fall off the edge' and disappear.
3. Normal: Individuals of age  $a$  are classified as ages which are normally distributed with mean  $a$  and constant c.v.  $c$ . As above, if there is no plus group in the population model, some individuals of the older age classes may disappear. If  $c$  is high enough, some of the younger age classes may 'fall off the other edge'. Individuals of age  $a < k$  are not misclassified.

Note that the expected values (fits) reported by CASAL<sup>2</sup> for observations with ageing error will have had the ageing error applied.

## 6.7. Simulating observations

CASAL<sup>2</sup> can generate simulated observations for a given model with given parameter values (using `spm -s`). Simulated observations are randomly distributed values, generated according to the error assumptions defined for each observation, around fits calculated from one or more sets of the 'true' parameter values. Simulating from a set of parameters can be used to generate observations from an operating model or as a form of parametric bootstrap.

The procedure CASAL<sup>2</sup> uses for simulating observations is to first run using the 'true' parameter values and generate the expected values. Then, if a set of observations uses ageing error, ageing error is applied. Finally a random value for each observed value is generated based on (i) the expected values, (ii) the type of likelihood specified, and (iii) the variability parameters (e.g., `error_value` and `process_error`).

Methods for generating the random error, and hence simulated values, depend on the specific likelihood type of each observation.

1. Normal likelihood parameterised by c.v.: Let  $E_i$  be the fitted value for observation  $i$ , and  $c_i$  be the corresponding c.v. (adjusted by the process error if applicable). Each simulated observation value  $S_i$  is generated as an independent normal deviate with mean  $E_i$  and standard deviation  $E_i c_i$ .

2. Log-normal likelihood: Let  $E_i$  be the fitted value for observation  $i$  and  $c_i$  be the corresponding c.v. (adjusted by the process error if applicable). Each simulated observation value  $S_i$  is generated as an independent lognormal deviate with mean and standard deviation (on the natural scale, not the log-scale) of  $E_i$  and  $E_i c_i$  respectively. The robustification parameter  $\delta$  is ignored.
3. Multinomial likelihood: Let  $E_i$  be the fitted value for observation  $i$ , for  $i$  between 1 and  $n$ , and let  $N$  be the sample size (adjusted by process error if applicable, and then rounded up to the next whole number). The robustification parameter  $\delta$  is ignored. Then,
  - a) A sample of  $N$  values from 1 to  $n$  is generated using the multinomial distribution, using sample probabilities proportional to the values of  $E_i$ .
  - b) Each simulated observation value  $S_i$  is calculated as the proportion of the  $N$  sampled values equalling  $i$
  - c) The simulated observation values  $S_i$  are then rescaled so that their sum is equal to 1
4. Binomial and the normal approximation to the binomial likelihoods: Let  $E_i$  be the fitted value for observation  $i$ , for  $i$  between 1 and  $n$ , and  $N_i$  the corresponding equivalent sample size (adjusted by process error if applicable, and then rounded up to the next whole number). The robustification parameter  $\delta$  is ignored. Then,
  - a) A sample of  $N_i$  independent binary variates is generated, equalling 1 with probability  $E_i$
  - b) The simulated observation value  $S_i$  is calculated as the sum of these binary variates divided by  $N_i$

Note that CASAL<sup>2</sup> will report simulated observations using the usual observation report (`@report[label].type=observation`). The report `@report[label].type=simulated.observation` will generate simulated observations in a form suitable for use as input within a CASAL<sup>2</sup> input configuration file. See Section 7 for more detail.

## 6.8. Pseudo-observations

CASAL<sup>2</sup> can generate expected values for observations without them contributing to the total objective function. These are called pseudo-observations, and can be used to either generate the expected values from CASAL<sup>2</sup> for reporting or diagnostic purposes. To define an observation as a pseudo-observation, use the command `@observation[label].likelihood=none`. Any observation type can be used as a pseudo-observation. CASAL<sup>2</sup> can also generate simulated observations from pseudo-observations. Note that;

- Output will only be generated if a report command `@report[label].type=observation` is specified.
- The observed values should be supplied (even if they are ‘dummy’ observation). These will be processed by CASAL<sup>2</sup> as if they were actual observation values, and must conform to the validations carried out for the other types of likelihood.
- The subcommands `likelihood`, `obs`, `error_value` and `process_error` have no effect when generating the expected values for the pseudo-observation.
- When simulating observations, CASAL<sup>2</sup> needs the subcommand `simulation.likelihood` to tell it what sort of likelihood to use. In this case, the `obs`, `error_value` and `process_error` are used to determine the appropriate terms to use for the likelihood when simulating.



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## 7. The report section

The report section specifies the printouts and other outputs from the model. CASAL<sup>2</sup> does not, in general, produce any output unless requested by a valid `@report` block.

Reports from CASAL<sup>2</sup> can be defined to print partition and states objects at a particular point in time, observation summaries, estimated parameters and objective function values. See below for a more extensive list, and an example of an observation report.

```
@report observation_age ## label of report
type observation ## Type of report
observation age_1990 ## label corresponding to an @observation report, shown below

@observation age_1990
type proportion_at_age
year 1990
plus_group
etc ...
```

Reports from CASAL<sup>2</sup> all conform to a standard style (with one exception — the `output_parameters` report, see below). The standard style is that reports are prefixed with an asterisk followed by a user-defined label and type of report in brackets (e.g., `*label (type)`), with the report ending with the line `*end`. For example,

```
*My_report (type)
...
*end
```

This syntax should make it easier for external packages to be configured to read CASAL<sup>2</sup> output. The `extract` functions in the **R** CASAL2 package uses this information to identify and read CASAL<sup>2</sup> output.

Note that the `output_parameters` report does not print either a header or `*end` at the end of the report. This is as the `output_parameters` report is designed to provide a single line (or multi-line for more than one set) vector of the estimated parameter values, suitable for reading by CASAL<sup>2</sup> (with the command `casal2 -i`). This is a specialised report for `casal2 -o` command. For estimate values in standard output users are recommended to use `type=estimate_value`.

Note that reports can be defined that may not be generated. For example printing the partition for a year and/or time-step that does not exist or reporting the covariance matrix when not estimating. Such reports are ignored by CASAL<sup>2</sup> and the program will not generate any output for these reports — although they must still conform to CASAL<sup>2</sup>'s syntax requirements.

Not all reports will be generated in all run modes. Some reports are only available in some run modes. For example, when simulating, only simulation reports will be output.

### 7.1. Print the partition

Print the partition for a given year or given years and time-step. This prints out, the numbers of individuals in each age class and category in the partition for each year. Note that this report is evaluated at the end of the time-step in the given year(s).

### 7.2. Print the partition at the end of an initialisation

Print the partition following an initialisation phase. This prints out, the numbers of individuals in each age class and category in the partition following an initialisation phase.

### 7.3. Print a process summary

Print a summary of a process. Depending on the process, different summaries are produced. These typically detail the type of process, its parameters and other options, and any associated details.

### 7.4. Print derived quantities

Print out the description of the derived quantity, and the values of the derived quantity as recorded in the model state, for each year of the model. and for all years in the initialisation phases.

### 7.5. Print the estimated parameters

Print a summary of the estimated parameters, including the parameter name, lower and upper bounds, the label of the prior, and its value.

### 7.6. Print the estimated parameters in a vector format

Print the estimated parameter values out as a vector. The `estimate_values` report prints the name of the parameter, followed by the value of that run.

### 7.7. Print the objective function

Print the total objective function value, and the value of all observations, the values of all priors, and the value of any penalties that have been incurred in the model. Note that if an individual model run does not incur a penalty, then the penalty will not be reported.

### 7.8. Print the covariance matrix

Print the Hessian and covariance matrices if estimating and if the covariance has been requested by `@minimiser[label].covariance=true`.

### 7.9. Print observations, fits, and residuals

Prints out for each category or combination of categories, expected values as calculated by the model, residuals (observed — expected), the error value, process error, and the total error (i.e., the error value as modified by any additional process error), and the contribution to the total objective function of that individual point in the observation.

Note that constants in likelihoods are often ignored in the objective function score of individual points. Hence, the total score from an observation equals the contribution of the objective function scores from each individual point plus a constant term (if applicable). In likelihoods without a constant term, then the total score from an observation will equal the contribution of the objective function scores from each individual point.

If simulating, then the contribution to the objective function of each observation is reported as zero.

### 7.10. Print simulated observations

Prints out a complete observation definition (i.e., in the form defined by `@report[label].type=observation`), but with observed values replaced by randomly generated simulated values. The output is in a form suitable for use within a CASAL<sup>2</sup> input configuration file, reproducing the command and subcommands from the input configuration file.



**7.11. Print the ageing error misclassification matrix**

Prints out the ageing error misclassification matrix.

**7.12. Print selectivities**

Prints the values of a selectivity for each age in the partition, for a given year and at then end of a given time-step.

**7.13. Print the random number seed**

Prints the random number seed used by CASAL<sup>2</sup> to generate the random number sequence. Future runs made with the same random number seed and the same model will produce identical outputs.

**7.14. Print the results of an MCMC**

Print the MCMC samples, objective function values, and proposal covariance matrix following an MCMC.

**7.15. Print the MCMC samples as they are calculated**

Print the MCMC samples for each new *i*th sample as they are calculated while doing an MCMC. The output file will be updated with each new sample as it is calculated by CASAL<sup>2</sup>.

**7.16. Print the MCMC objective function values as they are calculated**

Print the MCMC objective function values (along with the proposal covariance matrix) for each new *i*th sample as they are calculated while doing an MCMC. The output file will be updated with each new set of objective function values as it is calculated by CASAL<sup>2</sup>.

**7.17. Tabular reporting**

An alternative reporting framework to the standard output is the tabular reporting. Tabular reporting is used with multiline `-i` input files (like the MCMC reports). Tabular reports will print out a row that will correspond with each row of the `-i` input files. Tabular reporting is invoked at the command line using the following command `casal2 -r --tabular -i file_name`. Currently derived quantities and estimate values are the only report types that are within this framework. For each input file the output will begin with the names of each column followed by a multiline report ending with the `*end` syntax. These tables can be easily read into **R** using the CASAL2 package and for the example of MCMC multi-line files posteriors of derived quantities can be plotted.



---

## 8. Population command and subcommand syntax

For ease of reading CASAL<sup>2</sup> files in text editors, there exists a syntax highlighter `CASAL2.syn`

### 8.1. Model structure

`@model label` Define an object type Model

`age_plus` Define the oldest age as a plus group

Type: boolean

Default: false

Value: true, false

`final_year` Define the final year of the model, excluding years in the projection period

Type: non-negative integer

Default: No Default

Value: Defines the last year of the model, i.e., the model is run from `start_year` to `final_year`

`initialisation_phases` Define the labels of the phases of the initialisation

Type: string vector

Default: true

Value: A list of valid labels defined by `@initialisation_phase`

`label`

Type: string

Default: No Default

`length_bins`

Type: constant vector

Default: true

`max_age` Maximum age of individuals in the population

Type: non-negative integer

Default: 0

Value:  $0 \leq \text{age}_{\min} \leq \text{age}_{\max}$

`min_age` Minimum age of individuals in the population

Type: non-negative integer

Default: 0

Value:  $0 \leq \text{age}_{\min} \leq \text{age}_{\max}$

`projection_final_year` Define the final year of the model in projection mode

Type: non-negative integer

Default: 0

Value: Defines the last year of the projection period, i.e., the projection period runs from `final_year+1` to `projection_final_year`. For the default, 0, no projections are run.

**start\_year**      Define the first year of the model, immediately following initialisation  
Type: non-negative integer  
Default: No Default  
Value: Defines the first year of the model,  $\geq 1$ , e.g. 1990

**time\_steps**      Define the labels of the time steps, in the order that they are applied, to form the annual cycle  
Type: string vector  
Default: No Default  
Value: A list of valid labels defined by @time\_step

**type**      Type of model (the partition structure). Either age, length or hybrid  
Type: string  
Default: age

## 8.2. Initialisation

**@initialisation\_phase label**      Define an object type Initialisation Phase

**label**      Label  
Type: string  
Default: No Default

**type**      Type  
Type: string  
Default: iterative

### 8.2.1. @initialisation\_phase[label].type=cinitial

**categories**      List of categories to use  
Type: string vector  
Default: No Default

### 8.2.2. @initialisation\_phase[label].type=derived

**casal.initialisation.switch**      Reset the partition after running an extra annual cycle to take on equilibrium SSB's. Warning should only be set to true if comparing with previous CASAL models  
Type: boolean  
Default: false

**exclude\_processes**      The processes to exclude from all time steps  
Type: string vector  
Default: true

`insert_processes`      The processes to insert in to target time steps  
Type: string vector  
Default: true

### 8.2.3. `@initialisation_phase[label].type=iterative`

`convergence_years`      The years to test for convergence  
Type: non-negative integer vector  
Default: true

`exclude_processes`      The processes to exclude from all time steps  
Type: string vector  
Default: true

`insert_processes`      The processes to insert in to target time steps  
Type: string vector  
Default: true

`lambda`      Lambda  
Type: constant  
Default: Double(0.0)

`years`      The number of iterations to execute this phase for  
Type: non-negative integer  
Default: No Default

### 8.2.4. `@initialisation_phase[label].type=state_category_by_age`

`categories`      List of categories to use  
Type: string vector  
Default: No Default

`max_age`      Maximum age to use for this process  
Type: non-negative integer  
Default: No Default

`min_age`      Minimum age to use for this process  
Type: non-negative integer  
Default: No Default

### 8.3. Categories

**@categories** *label* Define an object type Categories

*age\_lengths* The labels of *age\_length* objects that are assigned to categories

Type: string vector

Default: true

*format* The format that the category names should adhere too

Type: string

Default: No Default

*names* The names of the categories to be used in the model

Type: string vector

Default: No Default

*years* The years that individual categories will be active for. This overrides the model values

Type: string vector

Default: true

### 8.4. Time-steps

**@time\_step** *label* Define an object type Time\_Step

*label* Label

Type: string

Default: No Default

*processes* Processes

Type: string vector

Default: No Default

*type*

Type: string

Default: No Default

### 8.5. Processes

**@process** *label* Define an object type Process

*print\_report* Generate parameter report

Type: boolean

Default: false

*label* Label

Type: string  
Default: No Default

type      Type  
Type: string  
Default: ""

### 8.5.1. @process[label].type=ageing

categories      Categories  
Type: string vector  
Default: No Default

print\_report      Generate parameter report  
Type: boolean  
Default: false

### 8.5.2. @process[label].type=growth

print\_report      Generate parameter report  
Type: boolean  
Default: false

### 8.5.3. @process[label].type=maturation

print\_report      Generate parameter report  
Type: boolean  
Default: false

from      List of categories to mature from  
Type: string vector  
Default: No Default

rates      The rates to mature for each year  
Type: constant vector  
Default: No Default

selectivities      List of selectivities to use for maturation  
Type: string vector  
Default: No Default

to      List of categories to mature too

Type: string vector  
Default: No Default

years      The years to be associated with rates  
Type: non-negative integer vector  
Default: No Default

#### **8.5.4. @process[label].type=mortality\_constant\_rate**

categories      List of categories  
Type: string vector  
Default: No Default

print\_report      Generate parameter report  
Type: boolean  
Default: false

m      Mortality rates  
Type: constant vector  
Default: No Default

time\_step\_ratio      Time step ratios for M  
Type: constant vector  
Default: true

selectivities      Selectivities  
Type: string vector  
Default: No Default

#### **8.5.5. @process[label].type=mortality\_event**

catches      Catches  
Type: constant vector  
Default: No Default

categories      Categories  
Type: string vector  
Default: No Default

print\_report      Generate parameter report  
Type: boolean  
Default: false

penalty      Penalty label



Type: string  
Default: ""

`selectivities`      List of selectivities  
Type: string vector  
Default: No Default

`u_max`      U Max  
Type: constant  
Default: 0.99

`years`      Years  
Type: non-negative integer vector  
Default: No Default

### 8.5.6. `@process[label].type=mortality_event_biomass`

`catches`      Catches for each year  
Type: constant vector  
Default: No Default

`categories`      Category labels  
Type: string vector  
Default: No Default

`print_report`      Generate parameter report  
Type: boolean  
Default: false

`penalty`      Penalty label  
Type: string  
Default: ""

`selectivities`      Selectivity labels  
Type: string vector  
Default: No Default

`u_max`      U Max  
Type: constant  
Default: 0.99

`units`      Unit of weight that the Catches table are expressed in  
Type: string  
Default: No Default

`years`      Years to apply mortality  
Type: non-negative integer vector  
Default: No Default

### 8.5.7. `@process[label].type=mortality holling rate`

a      parameter a  
Type: constant  
Default: No Default  
Lower Bound: 0.0 (inclusive)

b      parameter b  
Type: constant  
Default: No Default  
Lower Bound: 0.0 (inclusive)

`print_report`      Generate parameter report  
Type: boolean  
Default: false

`penalty`      Label of penalty to be applied  
Type: string  
Default: ""

`predator_categories`      Predator Categories labels  
Type: string vector  
Default: No Default

`predator_selectivities`      Selectivities for predator categories  
Type: string vector  
Default: No Default

`prey_categories`      Prey Categories labels  
Type: string vector  
Default: No Default

`prey_selectivities`      Selectivities for prey categories  
Type: string vector  
Default: No Default

`u_max`      Umax  
Type: constant  
Default: No Default  
Lower Bound: 0.0 (inclusive)  
Upper Bound: 1.0 (inclusive)

`x`      parameter x  
Type: constant  
Default: No Default  
Lower Bound: 0.0 (inclusive)

`years`      Year to execute in  
Type: non-negative integer vector  
Default: No Default

### 8.5.8. `@process[label].type=mortality_instantaneous`

`categories`      Categories for natural mortality  
Type: string vector  
Default: No Default

`print_report`      Generate parameter report  
Type: boolean  
Default: false

`m`      Mortality rates  
Type: constant vector  
Default: No Default  
Lower Bound: 0.0 (inclusive)  
Upper Bound: 1.0 (inclusive)

`selectivities`      Selectivities for Natural Mortality  
Type: string vector  
Default: No Default

`time_step_ratio`      Time step ratios for M  
Type: constant vector  
Default: true

`units`      Unit of weight that the Catches table are expressed in  
Type: string  
Default: No Default

**8.5.9. @process[label].type=mortality\_prey\_suitability**

consumption\_rate      Predator consumption rate

Type: constant

Default: No Default

Lower Bound: 0.0 (inclusive)

Upper Bound: 1.0 (inclusive)

print\_report      Generate parameter report

Type: boolean

Default: false

electivities      Prey Electivities

Type: constant vector

Default: No Default

Lower Bound: 0.0 (inclusive)

Upper Bound: 1.0 (inclusive)

penalty      Label of penalty to be applied

Type: string

Default: ""

predator\_categories      Predator Categories labels

Type: string vector

Default: No Default

predator\_selectivities      Selectivities for predator categories

Type: string vector

Default: No Default

prey\_categories      Prey Categories labels

Type: string vector

Default: No Default

prey\_selectivities      Selectivities for prey categories

Type: string vector

Default: No Default

u\_max      Umax

Type: constant

Default: No Default

Lower Bound: 0.0 (inclusive)

Upper Bound: 1.0 (inclusive)

years      Year that process occurs

Type: non-negative integer vector

Default: No Default

#### 8.5.10. `@process[label].type=nop`

`print_report`      Generate parameter report

Type: boolean

Default: false

#### 8.5.11. `@process[label].type=recruitment_beverton_holt`

`age`      Age to recruit at

Type: non-negative integer

Default: true

`b0`      B0

Type: constant

Default: false

`units`      Units of B0, if initialising model using B0

Type: string

Default: ""

`categories`      Category labels

Type: string vector

Default: No Default

`print_report`      Generate parameter report

Type: boolean

Default: false

`b0_initialisation_phase`      Initialisation phase Label that b0 is from

Type: string

Default: ""

`prior_standardised_ycs_values`      Priors for year class strength on ycs values (not standardised ycs values)

Type: boolean

Default: true

`proportions`      Proportions

Type: constant vector

Default: No Default

`r0`      **R0**

Type: constant

Default: false

`ssb`      **SSB Label (derived quantity**

Type: string

Default: No Default

`ssb_offset`      **Spawning biomass year offset**

Type: integer

Default: false

`standardise_ycs_years`      **Years that are included for year class standardisation**

Type: non-negative integer vector

Default: true

`steepness`      **Steepness**

Type: constant

Default: 1.0

`ycs_values`      **YCS Values**

Type: constant vector

Default: No Default

### **8.5.12. @process[label].type=recruitment\_constant**

`age`      **Age**

Type: non-negative integer

Default: No Default

`categories`      **Categories**

Type: string vector

Default: No Default

`print_report`      **Generate parameter report**

Type: boolean

Default: false

`proportions`      **Proportions**

Type: constant vector

Default: true

`r0`      **R0**

Type: constant  
Default: No Default  
Lower Bound: 0.0 (exclusive)

### 8.5.13. `@process[label].type=tag_by_age`

`print_report`      Generate parameter report  
Type: boolean  
Default: false

`from`      Categories to transition from  
Type: string vector  
Default: No Default

`initial_mortality`  
Type: constant  
Default: Double(0)

`initial_mortality_selectivity`  
Type: string  
Default: ""

`loss_rate`  
Type: constant vector  
Default: No Default

`loss_rate_selectivities`  
Type: string vector  
Default: true

`max_age`      Maximum age to transition  
Type: non-negative integer  
Default: No Default

`min_age`      Minimum age to transition  
Type: non-negative integer  
Default: No Default

`n`  
Type: constant vector  
Default: true

`penalty`      Penalty label  
Type: string  
Default: ""

`selectivities`

Type: string vector

Default: No Default

`to` Categories to transition to

Type: string vector

Default: No Default

`u_max` U Max

Type: constant

Default: 0.99

`years` Years to execute the transition in

Type: non-negative integer vector

Default: No Default

#### 8.5.14. `@process[label].type=tag_by_length`

`print_report` Generate parameter report

Type: boolean

Default: false

`from` Categories to transition from

Type: string vector

Default: No Default

`initial_mortality`

Type: constant

Default: Double(0)

`initial_mortality_selectivity`

Type: string

Default: ""

`maximum_length` The upper length when there is no plus group

Type: constant

Default: Double(0)

`n`

Type: constant vector

Default: true

`penalty` Penalty label



Type: string  
Default: ""

`plus_group`      Use plus group for last length bin  
Type: boolean  
Default: false

`selectivities`  
Type: string vector  
Default: No Default

`to`      Categories to transition to  
Type: string vector  
Default: No Default

`u_max`      U Max  
Type: constant  
Default: 0.99

`years`      Years to execute the transition in  
Type: non-negative integer vector  
Default: No Default

### 8.5.15. `@process[label].type=tag_loss`

`categories`      List of categories  
Type: string vector  
Default: No Default

`print_report`      Generate parameter report  
Type: boolean  
Default: false

`time_step_ratio`      Time step ratios for Tag Loss  
Type: constant vector  
Default: true

`selectivities`      Selectivities  
Type: string vector  
Default: No Default

`tag_loss_rate`      Tag Loss rates  
Type: constant vector  
Default: No Default

tag\_loss\_type    Type of tag loss  
Type: string  
Default: No Default

year    The year the first tagging release process was executed  
Type: non-negative integer  
Default: No Default

#### **8.5.16. @process[label].type=transition\_category**

print\_report    Generate parameter report  
Type: boolean  
Default: false

from    From  
Type: string vector  
Default: No Default

proportions    Proportions  
Type: constant vector  
Default: No Default

selectivities    Selectivity names  
Type: string vector  
Default: No Default

to    To  
Type: string vector  
Default: No Default

#### **8.5.17. @process[label].type=transition\_category\_by\_age**

print\_report    Generate parameter report  
Type: boolean  
Default: false

from    Categories to transition from  
Type: string vector  
Default: No Default

max\_age    Maximum age to transition  
Type: non-negative integer  
Default: No Default

`min_age` Minimum age to transition  
Type: non-negative integer  
Default: No Default

`penalty` Penalty label  
Type: string  
Default: ""

`to` Categories to transition to  
Type: string vector  
Default: No Default

`u_max` U Max  
Type: constant  
Default: 0.99

`years` Years to execute the transition in  
Type: non-negative integer vector  
Default: No Default

## 8.6. Time varying parameters

**@time\_varying** *label* Define an object type Time\_Varying

`label` Label  
Type: string  
Default: No Default

`parameter` Parameter to vary  
Type: string  
Default: No Default

`type` Type  
Type: string  
Default: ""

`years` Years to recalculate the values  
Type: non-negative integer vector  
Default: No Default

### 8.6.1. @time\_varying[label].type=annual\_shift

`a`  
Type: constant  
Default: No Default

b

Type: constant

Default: No Default

c

Type: constant

Default: No Default

parameter      Parameter to vary

Type: string

Default: No Default

scaling\_years

Type: non-negative integer vector

Default: true

values

Type: constant vector

Default: No Default

years      Years to recalculate the values

Type: non-negative integer vector

Default: No Default

### **8.6.2. @time\_varying[label].type=constant**

parameter      Parameter to vary

Type: string

Default: No Default

value      Value to assign to estimable

Type: constant

Default: No Default

years      Years to recalculate the values

Type: non-negative integer vector

Default: No Default

### **8.6.3. @time\_varying[label].type=exogenous**

a      Shift parameter

Type: constant

Default: No Default

`exogeneous_variable`      Values of exogeneous variable for each year  
Type: constant vector  
Default: No Default

`parameter`      Parameter to vary  
Type: string  
Default: No Default

`years`      Years to recalculate the values  
Type: non-negative integer vector  
Default: No Default

#### 8.6.4. `@time_varying[label].type=random_walk`

`distribution`      distribution  
Type: string  
Default: normal

`mean`      Mean  
Type: constant  
Default: 0

`parameter`      Parameter to vary  
Type: string  
Default: No Default

`sigma`      Standard deviation  
Type: constant  
Default: 1

`years`      Years to recalculate the values  
Type: non-negative integer vector  
Default: No Default

### 8.7. **Derived quantities**

`@derived_quantity label`      Define an object type `Derived.Quantity`

`categories`      The list of categories to use when calculating the derived quantity  
Type: string vector  
Default: No Default

`label`      Label

Type: string  
Default: No Default

time\_step\_proportion\_method  
Type: string  
Default: weighted\_sum  
Allowed Values: weighted\_sum, weighted\_product

selectivities      The list of selectivities to use when calculating the derived quantity. 1 per category  
Type: string vector  
Default: No Default

time\_step      The time step to calculate the derived quantity after  
Type: string  
Default: No Default

time\_step\_proportion  
Type: constant  
Default: Double(1.0)

type      Type  
Type: string  
Default: No Default

### 8.7.1. @derived\_quantity[label].type=abundance

categories      The list of categories to use when calculating the derived quantity  
Type: string vector  
Default: No Default

time\_step\_proportion\_method  
Type: string  
Default: weighted\_sum  
Allowed Values: weighted\_sum, weighted\_product

selectivities      The list of selectivities to use when calculating the derived quantity. 1 per category  
Type: string vector  
Default: No Default

time\_step      The time step to calculate the derived quantity after  
Type: string  
Default: No Default

time\_step\_proportion

Type: constant

Default: Double(1.0)

### 8.7.2. @derived\_quantity[label].type=biomass

categories      The list of categories to use when calculating the derived quantity

Type: string vector

Default: No Default

time\_step\_proportion\_method

Type: string

Default: weighted\_sum

Allowed Values: weighted\_sum, weighted\_product

selectivities      The list of selectivities to use when calculating the derived quantity. 1 per category

Type: string vector

Default: No Default

time\_step      The time step to calculate the derived quantity after

Type: string

Default: No Default

time\_step\_proportion

Type: constant

Default: Double(1.0)

## 8.8. Age-length relationship

**@age\_length** *label*      Define an object type Age\_Length

casal\_switch      A switch to use CASAL Cumulative normal function, note CASAL2 uses the recent BOOST function which differs from the previous CASAL algorithm

Type: boolean

Default: false

cv\_first      CV for the first age class

Type: constant

Default: Double(0.0

Lower Bound: 0.0 (inclusive)

cv\_last      CV for last age class

Type: constant  
Default: Double(0.0)  
Lower Bound: 0.0 (inclusive)

distribution      TBA  
Type: string  
Default: normal

label      Label  
Type: string  
Default: No Default

time\_step\_proportions      the proportion increase of age through the in each time step that corresponds to a length and thus weight increase  
Type: constant vector  
Default: true

type      Type  
Type: string  
Default: No Default

### 8.8.1. @age\_\_length[label].type=data

by\_length      Specifies if the linear interpolation of CV's is a linear function of mean length at age. Default is just by age  
Type: boolean  
Default: true

casal\_switch      A switch to use CASAL Cumulative normal function, note CASAL2 uses the recent BOOST function which differs from the previous CASAL algorithm  
Type: boolean  
Default: false

cv\_first      CV for the first age class  
Type: constant  
Default: Double(0.0)  
Lower Bound: 0.0 (inclusive)

cv\_last      CV for last age class  
Type: constant  
Default: Double(0.0)  
Lower Bound: 0.0 (inclusive)

distribution      TBA



Type: string  
Default: normal

external\_gaps  
Type: string  
Default: mean  
Allowed Values: mean, nearest\_neighbour

internal\_gaps  
Type: string  
Default: mean  
Allowed Values: mean, nearest\_neighbour, interpolate

length\_weight      TBA  
Type: string  
Default: No Default

time\_step\_proportions      the proportion increase of age through the in each time step that corresponds to a length and thus weight increase  
Type: constant vector  
Default: true

### 8.8.2. @age\_length[label].type=none

casal\_switch      A switch to use CASAL Cumulative normal function, note CASAL2 uses the recent BOOST function which differs from the previous CASAL algorithm  
Type: boolean  
Default: false

cv\_first      CV for the first age class  
Type: constant  
Default: Double(0.0  
Lower Bound: 0.0 (inclusive)

cv\_last      CV for last age class  
Type: constant  
Default: Double(0.0  
Lower Bound: 0.0 (inclusive)

distribution      TBA  
Type: string  
Default: normal

time\_step\_proportions      the proportion increase of age through the in each time step that

corresponds to a length and thus weight increase

Type: constant vector

Default: true

### 8.8.3. `@age__length[label].type=schnute`

a TBA

Type: constant

Default: No Default

Lower Bound: 0.0 (inclusive)

b TBA

Type: constant

Default: No Default

Lower Bound: 0.0 (exclusive)

by\_length TBA

Type: boolean

Default: true

casal\_switch A switch to use CASAL Cumulative normal function, note CASAL2 uses the recent BOOST function which differs from the previous CASAL algorithm

Type: boolean

Default: false

cv\_first CV for the first age class

Type: constant

Default: Double(0.0

Lower Bound: 0.0 (inclusive)

cv\_last CV for last age class

Type: constant

Default: Double(0.0

Lower Bound: 0.0 (inclusive)

distribution TBA

Type: string

Default: normal

length\_weight TBA

Type: string

Default: No Default

taul TBA

Type: constant  
Default: No Default

tau2      TBA  
Type: constant  
Default: No Default

time\_step\_proportions      the proportion increase of age through the in each time step that corresponds to a length and thus weight increase  
Type: constant vector  
Default: true

y1      TBA  
Type: constant  
Default: No Default

y2      TBA  
Type: constant  
Default: No Default

#### 8.8.4. @age\_length[label].type=von bertalanffy

by\_length      Specifies if the linear interpolation of CV's is a linear function of mean length at age. Default is just by age  
Type: boolean  
Default: true

casal\_switch      A switch to use CASAL Cumulative normal function, note CASAL2 uses the recent BOOST function which differs from the previous CASAL algorithm  
Type: boolean  
Default: false

cv\_first      CV for the first age class  
Type: constant  
Default: Double(0.0  
Lower Bound: 0.0 (inclusive)

cv\_last      CV for last age class  
Type: constant  
Default: Double(0.0  
Lower Bound: 0.0 (inclusive)

distribution      TBA  
Type: string  
Default: normal

**k**      **TBA**

Type: constant

Default: No Default

Lower Bound: 0.0 (inclusive)

**length\_weight**      **TBA**

Type: string

Default: No Default

**linf**      **TBA**

Type: constant

Default: No Default

Lower Bound: 0.0 (inclusive)

**t0**      **TBA**

Type: constant

Default: No Default

**time\_step\_proportions**      the proportion increase of age through the in each time step that corresponds to a length and thus weight increase

Type: constant vector

Default: true

## 8.9. Length-weight

**@length\_weight** *label*      Define an object type Length\_Weight

**label**      Label

Type: string

Default: No Default

**type**      Type

Type: string

Default: No Default

### 8.9.1. @length\_weight[label].type=basic

**a**      **A**

Type: constant

Default: No Default

**b**      **B**

Type: constant

Default: No Default

`units`      Units of measure (tonnes, kgs, grams)  
Type: string  
Default: No Default

### 8.9.2. `@length_weight[label].type=none`

## 8.10. Selectivities

**`@selectivity label`**      Define an object type Selectivity

`label`      Label  
Type: string  
Default: No Default

`length_based`      Is the selectivity length based  
Type: boolean  
Default: false

`intervals`      Number of quantiles to evaluate a length based selectivity over the age length distribution  
Type: non-negative integer  
Default: 5

`type`      Type  
Type: string  
Default: No Default

### 8.10.1. `@selectivity[label].type=all_values`

`length_based`      Is the selectivity length based  
Type: boolean  
Default: false

`intervals`      Number of quantiles to evaluate a length based selectivity over the age length distribution  
Type: non-negative integer  
Default: 5

`v`      V  
Type: constant vector  
Default: No Default

### 8.10.2. @selectivity[label].type=all\_values\_bounded

h H

Type: non-negative integer

Default: No Default

length\_based Is the selectivity length based

Type: boolean

Default: false

l L

Type: non-negative integer

Default: No Default

intervals Number of quantiles to evaluate a length based selectivity over the age length distribution

Type: non-negative integer

Default: 5

v V

Type: constant vector

Default: No Default

### 8.10.3. @selectivity[label].type=constant

c C

Type: constant

Default: No Default

length\_based Is the selectivity length based

Type: boolean

Default: false

intervals Number of quantiles to evaluate a length based selectivity over the age length distribution

Type: non-negative integer

Default: 5

### 8.10.4. @selectivity[label].type=double\_exponential

alpha Alpha

Type: constant

Default: 1.0

`length_based`     Is the selectivity length based  
Type: boolean  
Default: false

`intervals`     Number of quantiles to evaluate a length based selectivity over the age length distribution  
Type: non-negative integer  
Default: 5

`x0`     **X0**  
Type: constant  
Default: No Default

`x1`     **X1**  
Type: constant  
Default: No Default

`x2`     **X2**  
Type: constant  
Default: No Default

`y0`     **Y0**  
Type: constant  
Default: No Default

`y1`     **Y1**  
Type: constant  
Default: No Default

`y2`     **Y2**  
Type: constant  
Default: No Default

#### **8.10.5. @selectivity[label].type=double\_normal**

`alpha`     **Alpha**  
Type: constant  
Default: 1.0

`length_based`     Is the selectivity length based  
Type: boolean  
Default: false

`mu`     **Mu**

Type: constant  
Default: No Default

`intervals`      Number of quantiles to evaluate a length based selectivity over the age length distribution  
Type: non-negative integer  
Default: 5

`sigma_l`      Sigma L  
Type: constant  
Default: No Default

`sigma_r`      Sigma R  
Type: constant  
Default: No Default

### 8.10.6. `@selectivity[label].type=increasing`

`alpha`      Alpha  
Type: constant  
Default: 1.0

`h`      High  
Type: non-negative integer  
Default: No Default

`length_based`      Is the selectivity length based  
Type: boolean  
Default: false

`l`      Low  
Type: non-negative integer  
Default: No Default

`intervals`      Number of quantiles to evaluate a length based selectivity over the age length distribution  
Type: non-negative integer  
Default: 5

`v`      V  
Type: constant vector  
Default: No Default



**8.10.7. @selectivity[label].type=inverse\_logistic**

a50      A50  
Type: constant  
Default: No Default

alpha     Alpha  
Type: constant  
Default: 1.0

ato95     aTo95  
Type: constant  
Default: No Default

length\_based    Is the selectivity length based  
Type: boolean  
Default: false

intervals      Number of quantiles to evaluate a length based selectivity over the age length  
distribution  
Type: non-negative integer  
Default: 5

**8.10.8. @selectivity[label].type=knife\_edge**

alpha     Alpha  
Type: constant  
Default: 1.0

e          Edge  
Type: constant  
Default: No Default

length\_based    Is the selectivity length based  
Type: boolean  
Default: false

intervals      Number of quantiles to evaluate a length based selectivity over the age length  
distribution  
Type: non-negative integer  
Default: 5

### 8.10.9. **@selectivity[label].type=logistic**

a50      A50

Type: constant

Default: No Default

alpha     Alpha

Type: constant

Default: 1.0

ato95     Ato95

Type: constant

Default: No Default

length\_based    Is the selectivity length based

Type: boolean

Default: false

intervals       Number of quantiles to evaluate a length based selectivity over the age length  
distribution

Type: non-negative integer

Default: 5

### 8.10.10. **@selectivity[label].type=logistic\_producing**

a50      A50

Type: constant

Default: No Default

alpha     Alpha

Type: constant

Default: 1.0

ato95     Ato95

Type: constant

Default: No Default

h          High

Type: non-negative integer

Default: No Default

length\_based    Is the selectivity length based

Type: boolean

Default: false

---

`l`      **Low**  
Type: non-negative integer  
Default: No Default

`intervals`      Number of quantiles to evaluate a length based selectivity over the age length distribution  
Type: non-negative integer  
Default: 5

## 9. Estimation command and subcommand syntax

### 9.1. Estimation methods

**@estimate** *label*      Define an object type Estimate

`estimation_phase`      TBA  
Type: non-negative integer  
Default: 1u

`label`      **Label**  
Type: string  
Default: ""

`lower_bound`      The lowest value the parameter is allowed to have  
Type: constant  
Default: No Default

`mcmc`      TBA  
Type: boolean  
Default: false

`parameter`      The name of the variable to estimate in the model  
Type: string  
Default: No Default

`prior`      The name of the prior to use for the parameter  
Type: string  
Default: ""

`same`      A list of parameters that are bound to the value of this estimate  
Type: string vector  
Default: ""

`type`      **Type**

Type: string  
Default: No Default

`upper_bound`      The highest value the parameter is allowed to have  
Type: constant  
Default: No Default

### 9.1.1. `@estimate[label].type=beta`

a      A  
Type: constant  
Default: No Default

b      B  
Type: constant  
Default: No Default

`estimation_phase`      TBA  
Type: non-negative integer  
Default: 1u

`lower_bound`      The lowest value the parameter is allowed to have  
Type: constant  
Default: No Default

`mcmc`      TBA  
Type: boolean  
Default: false

`mu`      Mu  
Type: constant  
Default: No Default

`parameter`      The name of the variable to estimate in the model  
Type: string  
Default: No Default

`prior`      The name of the prior to use for the parameter  
Type: string  
Default: ""

`same`      A list of parameters that are bound to the value of this estimate  
Type: string vector  
Default: ""

`sigma`      **Sigma**

Type: constant

Default: No Default

Lower Bound: 0.0 (exclusive)

`upper_bound`      The highest value the parameter is allowed to have

Type: constant

Default: No Default

### 9.1.2. `@estimate[label].type=lognormal`

`cv`      **Cv**

Type: constant

Default: No Default

Lower Bound: 0.0 (exclusive)

`estimation_phase`      **TBA**

Type: non-negative integer

Default: 1u

`lower_bound`      The lowest value the parameter is allowed to have

Type: constant

Default: No Default

`mcmc`      **TBA**

Type: boolean

Default: false

`mu`      **Mu**

Type: constant

Default: No Default

Lower Bound: 0.0 (exclusive)

`parameter`      The name of the variable to estimate in the model

Type: string

Default: No Default

`prior`      The name of the prior to use for the parameter

Type: string

Default: ""

`same`      A list of parameters that are bound to the value of this estimate

Type: string vector

Default: ""

`upper_bound`     The highest value the parameter is allowed to have  
Type: constant  
Default: No Default

### 9.1.3. `@estimate[label].type=normal`

`cv`     `Cv`  
Type: constant  
Default: No Default  
Lower Bound: 0.0 (exclusive)

`estimation_phase`     TBA  
Type: non-negative integer  
Default: 1u

`lower_bound`     The lowest value the parameter is allowed to have  
Type: constant  
Default: No Default

`mcmc`     TBA  
Type: boolean  
Default: false

`mu`     `Mu`  
Type: constant  
Default: No Default

`parameter`     The name of the variable to estimate in the model  
Type: string  
Default: No Default

`prior`     The name of the prior to use for the parameter  
Type: string  
Default: ""

`same`     A list of parameters that are bound to the value of this estimate  
Type: string vector  
Default: ""

`upper_bound`     The highest value the parameter is allowed to have  
Type: constant  
Default: No Default

**9.1.4. @estimate[label].type=normal\_by\_stdev**

estimation\_phase    TBA

Type: non-negative integer

Default: 1u

lower\_bound    The lowest value the parameter is allowed to have

Type: constant

Default: No Default

mcmc    TBA

Type: boolean

Default: false

mu    Mu

Type: constant

Default: No Default

parameter    The name of the variable to estimate in the model

Type: string

Default: No Default

prior    The name of the prior to use for the parameter

Type: string

Default: ""

same    A list of parameters that are bound to the value of this estimate

Type: string vector

Default: ""

sigma    Sigma

Type: constant

Default: No Default

Lower Bound: 0.0 (exclusive)

upper\_bound    The highest value the parameter is allowed to have

Type: constant

Default: No Default

**9.1.5. @estimate[label].type=normal\_log**

estimation\_phase    TBA

Type: non-negative integer

Default: 1u

`lower_bound`     The lowest value the parameter is allowed to have  
Type: constant  
Default: No Default

`mcmc`     TBA  
Type: boolean  
Default: false

`mu`     Mu  
Type: constant  
Default: No Default

`parameter`     The name of the variable to estimate in the model  
Type: string  
Default: No Default

`prior`     The name of the prior to use for the parameter  
Type: string  
Default: ""

`same`     A list of parameters that are bound to the value of this estimate  
Type: string vector  
Default: ""

`sigma`     Sigma  
Type: constant  
Default: No Default  
Lower Bound: 0.0 (exclusive)

`upper_bound`     The highest value the parameter is allowed to have  
Type: constant  
Default: No Default

### 9.1.6. `@estimate[label].type=uniform`

`estimation_phase`     TBA  
Type: non-negative integer  
Default: 1u

`lower_bound`     The lowest value the parameter is allowed to have  
Type: constant  
Default: No Default

`mcmc`     TBA



Type: boolean

Default: false

`parameter`     The name of the variable to estimate in the model

Type: string

Default: No Default

`prior`     The name of the prior to use for the parameter

Type: string

Default: ""

`same`     A list of parameters that are bound to the value of this estimate

Type: string vector

Default: ""

`upper_bound`     The highest value the parameter is allowed to have

Type: constant

Default: No Default

### 9.1.7. `@estimate[label].type=uniform_log`

`estimation_phase`     TBA

Type: non-negative integer

Default: 1u

`lower_bound`     The lowest value the parameter is allowed to have

Type: constant

Default: No Default

`mcmc`     TBA

Type: boolean

Default: false

`parameter`     The name of the variable to estimate in the model

Type: string

Default: No Default

`prior`     The name of the prior to use for the parameter

Type: string

Default: ""

`same`     A list of parameters that are bound to the value of this estimate

Type: string vector

Default: ""

`upper_bound`      The highest value the parameter is allowed to have  
     Type: constant  
     Default: No Default

## 9.2. Point estimation

**@`minimiser` `label`**      Define an object type `Minimiser`

`active`      True if this minimiser is active  
     Type: boolean  
     Default: false

`covariance`      True if a covariance matrix should be created  
     Type: boolean  
     Default: true

`label`      Label  
     Type: string  
     Default: No Default

`type`      Type of minimiser to use  
     Type: string  
     Default: No Default

### 9.2.1. @`minimiser`[`label`].`type=callback_a_d_o_l_c`

`active`      True if this minimiser is active  
     Type: boolean  
     Default: false

`covariance`      True if a covariance matrix should be created  
     Type: boolean  
     Default: true

`tolerance`      Tolerance of the gradient for convergence  
     Type: constant  
     Default: 0.02

`evaluations`      Maximum number of evaluations  
     Type: integer  
     Default: 4000

`iterations`      Maximum number of iterations  
     Type: integer  
     Default: 1000

`step_size` Minimum Step-size before minimisation fails  
Type: constant  
Default: 1e-7

### 9.2.2. `@minimiser[label].type=engine_a_d_o_l_c`

`active` True if this minimiser is active  
Type: boolean  
Default: false

`covariance` True if a covariance matrix should be created  
Type: boolean  
Default: true

`tolerance` Tolerance of the gradient for convergence  
Type: constant  
Default: 0.02

`evaluations` Maximum number of evaluations  
Type: integer  
Default: 4000

`iterations` Maximum number of iterations  
Type: integer  
Default: 1000

`step_size` Minimum Step-size before minimisation fails  
Type: constant  
Default: 1e-7

### 9.2.3. `@minimiser[label].type=f_m_m_a_d_o_l_c`

`active` True if this minimiser is active  
Type: boolean  
Default: false

`covariance` True if a covariance matrix should be created  
Type: boolean  
Default: true

`tolerance` Tolerance of the gradient for convergence  
Type: constant  
Default: 0.02

evaluations      Maximum number of evaluations  
Type: integer  
Default: 4000

iterations      Maximum number of iterations  
Type: integer  
Default: 1000

step\_size      Minimum Step-size before minimisation fails  
Type: constant  
Default: 1e-7

#### **9.2.4. @minimiser[label].type=beta\_diff**

active      True if this minimiser is active  
Type: boolean  
Default: false

covariance      True if a covariance matrix should be created  
Type: boolean  
Default: true

tolerance      Tolerance of the gradient for convergence  
Type: constant  
Default: 2e-3

evaluations      Maximum number of evaluations  
Type: integer  
Default: 4000

iterations      Maximum number of iterations  
Type: integer  
Default: 1000

#### **9.2.5. @minimiser[label].type=c\_p\_p\_a\_d**

active      True if this minimiser is active  
Type: boolean  
Default: false

covariance      True if a covariance matrix should be created  
Type: boolean  
Default: true

**9.2.6. @minimiser[label].type=call\_back\_de\_solver**

active      True if this minimiser is active

Type: boolean

Default: false

covariance      True if a covariance matrix should be created

Type: boolean

Default: true

crossover\_probability      TBA

Type: constant

Default: 0.9

difference\_scale      The scale to apply to new solutions when comparing candidates

Type: constant

Default: 0.02

max\_generations      The maximum number of iterations to run

Type: non-negative integer

Default: No Default

method      The type of candidate generation method to use

Type: string

Default: ""

Value: not\_yet\_implemented

population\_size      The number of candidate solutions to have in the population

Type: non-negative integer

Default: No Default

tolerance      The total variance between the population and best candidate before acceptance

Type: constant

Default: 0.01

**9.2.7. @minimiser[label].type=engine\_de\_solver**

active      True if this minimiser is active

Type: boolean

Default: false

covariance      True if a covariance matrix should be created

Type: boolean  
Default: true

crossover\_probability      TBA  
Type: constant  
Default: 0.9

difference\_scale      The scale to apply to new solutions when comparing candidates  
Type: constant  
Default: 0.02

max\_generations      The maximum number of iterations to run  
Type: non-negative integer  
Default: No Default

method      The type of candidate generation method to use  
Type: string  
Default: ""  
Value: not\_yet\_implemented

### 9.2.8. @minimiser[label].type=call\_back\_dlib

active      True if this minimiser is active  
Type: boolean  
Default: false

covariance      True if a covariance matrix should be created  
Type: boolean  
Default: true

### 9.2.9. @minimiser[label].type=dummy

active      True if this minimiser is active  
Type: boolean  
Default: false

covariance      True if a covariance matrix should be created  
Type: boolean  
Default: true

**9.2.10. @minimiser[label].type=callback\_gamma\_diff**

active	True if this minimiser is active
Type: boolean	
Default: false	
covariance	True if a covariance matrix should be created
Type: boolean	
Default: true	
tolerance	Tolerance of the gradient for convergence
Type: constant	
Default: 0.02	
evaluations	Maximum number of evaluations
Type: integer	
Default: 4000	
iterations	Maximum number of iterations
Type: integer	
Default: 1000	
step_size	Minimum Step-size before minimisation fails
Type: constant	
Default: 1e-7	

**9.2.11. @minimiser[label].type=engine\_gamma\_diff**

active	True if this minimiser is active
Type: boolean	
Default: false	
covariance	True if a covariance matrix should be created
Type: boolean	
Default: true	
tolerance	Tolerance of the gradient for convergence
Type: constant	
Default: 0.02	
evaluations	Maximum number of evaluations
Type: integer	
Default: 4000	
iterations	Maximum number of iterations

Type: integer  
Default: 1000

`step_size` Minimum Step-size before minimisation fails  
Type: constant  
Default: 1e-7

### 9.2.12. `@minimiser[label].type=f m m_gamma_diff`

`active` True if this minimiser is active  
Type: boolean  
Default: false

`covariance` True if a covariance matrix should be created  
Type: boolean  
Default: true

`tolerance` Tolerance of the gradient for convergence  
Type: constant  
Default: 0.02

`evaluations` Maximum number of evaluations  
Type: integer  
Default: 4000

`iterations` Maximum number of iterations  
Type: integer  
Default: 1000

`step_size` Minimum Step-size before minimisation fails  
Type: constant  
Default: 1e-7

## 9.3. Monte Carlo Markov Chain (MCMC)

`@mcmc label` Define an object type MCMC

`active` Is this the active MCMC algorithm  
Type: boolean  
Default: true

`label` Label  
Type: string  
Default: No Default



`length`      The number of chain links to create  
Type: non-negative integer  
Default: No Default

`print_default_reports`  
Type: boolean  
Default: true

`type`      Type  
Type: string  
Default: ""

### 9.3.1. `@m.c.m.c[label].type=independence metropolis`

`active`      Is this the active MCMC algorithm  
Type: boolean  
Default: true

`adapt_stepsize_at`      Iterations in the chain to check and resize the MCMC stepsize  
Type: non-negative integer vector  
Default: true

`correlation_adjustment_diff`      TBA  
Type: constant  
Default: 0.0001

`covariance_adjustment_method`      Method for adjusting small variances in the covariance proposal matrix  
Type: string  
Default: covariance

`df`      Degrees of freedom of the multivariate t proposal distribution  
Type: non-negative integer  
Default: 4

`keep`      Spacing between recorded values in the chain  
Type: non-negative integer  
Default: 1u

`length`      The number of chain links to create  
Type: non-negative integer  
Default: No Default

`max_correlation`      Maximum absolute correlation in the covariance matrix of the proposal

**distribution**  
Type: constant  
Default: 0.8

**print\_default\_reports**  
Type: boolean  
Default: true

**proposal\_distribution**      The shape of the proposal distribution (either t or normal)  
Type: string  
Default: t

**start**      Covariance multiplier for the starting point of the Markov chain  
Type: constant  
Default: 0.0

**step\_size**      Initial stepsize (as a multiplier of the approximate covariance matrix)  
Type: constant  
Default: 0.02

## 9.4. Profiles

**@profile** *label*      Define an object type Profile

**label**      Label  
Type: string  
Default: ""

**lower\_bound**      The lower bounds  
Type: constant  
Default: No Default

**parameter**      The system parameter to profile  
Type: string  
Default: No Default

**steps**      The number of steps to take between the lower and upper bound  
Type: non-negative integer  
Default: No Default

**type**  
Type: string  
Default: No Default

**upper\_bound**      The upper bounds

Type: constant  
Default: No Default

## 9.5. Defining catchability constants

**@catchability** *label*     Define an object type Catchability

*label*     Label  
Type: string  
Default: No Default

*type*  
Type: string  
Default: No Default

### 9.5.1. @catchability[label].type=free

*q*     The catchability amount  
Type: constant  
Default: No Default

## 9.6. Defining penalties

**@penalty** *label*     Define an object type Penalty

*label*     Label  
Type: string  
Default: No Default

*type*     Type  
Type: string  
Default: No Default

### 9.6.1. @penalty[label].type=process

*log\_scale*     Log scale  
Type: boolean  
Default: false

*multiplier*     Multiplier  
Type: constant  
Default: 1.0

## 9.7. Defining priors on parameter ratios, differences and means

**@additional\_prior** *label* Define an object type Additional\_Prior

*label*      **Label**  
Type: string  
Default: No Default

*type*      **Type**  
Type: string  
Default: No Default

### 9.7.1. @additional\_\_prior[label].type=beta

*a*      **A**  
Type: constant  
Default: No Default

*b*      **B**  
Type: constant  
Default: No Default

*mu*      **Mu**  
Type: constant  
Default: No Default

*sigma*      **Sigma**  
Type: constant  
Default: No Default  
Lower Bound: 0.0 (inclusive)

### 9.7.2. @additional\_\_prior[label].type=vector\_average

*k*      **K Value to use in the calculation**  
Type: constant  
Default: No Default

*method*      **What calculation method to use (k, l, m**  
Type: string  
Default: k

*multiplier*      **Multiplier for the penalty amount**  
Type: constant  
Default: 1

---

parameter      Label of the estimate to generate penalty on  
Type: string  
Default: No Default

### 9.7.3. @additional\_prior[label].type=vector\_smoothing

log\_scale      Log scale  
Type: boolean  
Default: false

lower\_bound      First element to apply the penalty to in the vector  
Type: non-negative integer  
Default: 0u

multiplier      Multiplier for the penalty amount  
Type: constant  
Default: 1

parameter      Label of the estimate to generate penalty on  
Type: string  
Default: No Default

r      Penalty applied to rth differences  
Type: non-negative integer  
Default: 2u

upper\_bound      Last element to apply the penalty to in the vector  
Type: non-negative integer  
Default: 0u

## 10. Observation command and subcommand syntax

### 10.1. Observation types

The observation types available are,

Observations of proportions of individuals by age class

Observations of proportions of individuals between categories within each age class

Relative and absolute abundance observations

Relative and absolute biomass observations

Each type of observation requires a set of subcommands and arguments specific to that process.

**@observation** *label*      Define an object type Observation

`categories`      Category labels to use  
Type: string vector  
Default: true

`error_value_multiplier`      Error value multiplier for likelihood  
Type: constant  
Default: Double(1.0)

`label`      Label  
Type: string  
Default: No Default

`likelihood_multiplier`      Likelihood score multiplier  
Type: constant  
Default: Double(1.0)

`likelihood`      Type of likelihood to use  
Type: string  
Default: No Default

`simulation_likelihood`      Simulation likelihood to use  
Type: string  
Default: ""

`type`      Type of observation  
Type: string  
Default: No Default

### 10.1.1. `@observation[label].type=process_abundance`

`catchability`      Abundance catchability  
Type: string  
Default: No Default

`categories`      Category labels to use  
Type: string vector  
Default: true

`delta`      Delta value for error values  
Type: constant  
Default: Double(1e-10)

`error_value_multiplier`      Error value multiplier for likelihood  
Type: constant  
Default: Double(1.0)

`error_value`      The error values to use against the observation values  
Type: constant vector  
Default: No Default

`likelihood_multiplier`      Likelihood score multiplier  
Type: constant  
Default: Double(1.0)

`likelihood`      Type of likelihood to use  
Type: string  
Default: No Default

`obs`      Observation values  
Type: string vector  
Default: No Default

`process_error`      Process error  
Type: constant  
Default: Double(0.0)

`process`      Process label  
Type: string  
Default: No Default

`process_proportion`      Process proportion  
Type: constant  
Default: Double(0.5)

`selectivities`      Selectivity labels to use  
Type: string vector  
Default: true

`simulation_likelihood`      Simulation likelihood to use  
Type: string  
Default: ""

`time_step`      Time step to execute in  
Type: string  
Default: No Default

`years`      Years to execute in  
Type: non-negative integer vector  
Default: No Default

**10.1.2. @observation[label].type=time\_step\_abundance**

catchability      TBA

Type: string

Default: No Default

categories      Category labels to use

Type: string vector

Default: true

delta      Delta value for error values

Type: constant

Default: Double(1e-10)

error\_value\_multiplier      Error value multiplier for likelihood

Type: constant

Default: Double(1.0)

error\_value      The error values to use against the observation values

Type: constant vector

Default: No Default

likelihood\_multiplier      Likelihood score multiplier

Type: constant

Default: Double(1.0)

likelihood      Type of likelihood to use

Type: string

Default: No Default

obs      Observation values

Type: string vector

Default: No Default

process\_error      Process error

Type: constant

Default: Double(0.0)

selectivities      Selectivity labels to use

Type: string vector

Default: true

simulation\_likelihood      Simulation likelihood to use

Type: string

Default: ""



`time_step`      Time step to execute in  
Type: string  
Default: No Default

`time_step_proportion`      Proportion through the time step to analyse the partition from  
Type: constant  
Default: Double(0.5)

`years`      Years to execute in  
Type: non-negative integer vector  
Default: No Default

### 10.1.3. `@observation[label].type=process_biomass`

`catchability`      Catchability of Biomass  
Type: string  
Default: No Default

`categories`      Category labels to use  
Type: string vector  
Default: true

`delta`      Delta value for error values  
Type: constant  
Default: Double(1e-10)

`error_value_multiplier`      Error value multiplier for likelihood  
Type: constant  
Default: Double(1.0)

`error_value`      The error values to use against the observation values  
Type: constant vector  
Default: No Default

`likelihood_multiplier`      Likelihood score multiplier  
Type: constant  
Default: Double(1.0)

`likelihood`      Type of likelihood to use  
Type: string  
Default: No Default

`obs`      Observation values  
Type: string vector  
Default: No Default

`process_error`      Process error  
Type: constant  
Default: Double(0.0)

`process`      Process label  
Type: string  
Default: No Default

`process_proportion`      Process proportion  
Type: constant  
Default: Double(0.5)

`selectivities`      Selectivity labels to use  
Type: string vector  
Default: true

`simulation_likelihood`      Simulation likelihood to use  
Type: string  
Default: ""

`time_step`      Time step to execute in  
Type: string  
Default: No Default

`years`      Years to execute in  
Type: non-negative integer vector  
Default: No Default

#### 10.1.4. `@observation[label].type=time_step_biomass`

`catchability`      Catchability of Biomass  
Type: string  
Default: No Default

`categories`      Category labels to use  
Type: string vector  
Default: true

`delta`      Delta value for error values  
Type: constant  
Default: Double(1e-10)

`error_value_multiplier`      Error value multiplier for likelihood

	Type: constant Default: Double(1.0)
error_value	The error values to use against the observation values Type: constant vector Default: No Default
likelihood_multiplier	Likelihood score multiplier Type: constant Default: Double(1.0)
likelihood	Type of likelihood to use Type: string Default: No Default
obs	Observation values Type: string vector Default: No Default
process_error	Process error Type: constant Default: Double(0.0)
selectivities	Selectivity labels to use Type: string vector Default: true
simulation_likelihood	Simulation likelihood to use Type: string Default: ""
time_step	Time step to execute in Type: string Default: No Default
time_step_proportion	Proportion through the time step to analyse the partition from Type: constant Default: Double(0.5)
years	Years to execute in Type: non-negative integer vector Default: No Default

**10.1.5. @observation[label].type=process\_proportions\_at\_age**

age\_plus      Use age plus group

Type: boolean

Default: true

ageing\_error      Label of ageing error to use

Type: string

Default: ""

categories      Category labels to use

Type: string vector

Default: true

delta      Delta

Type: constant

Default: DELTA

error\_value\_multiplier      Error value multiplier for likelihood

Type: constant

Default: Double(1.0)

likelihood\_multiplier      Likelihood score multiplier

Type: constant

Default: Double(1.0)

likelihood      Type of likelihood to use

Type: string

Default: No Default

max\_age      Maximum age

Type: non-negative integer

Default: No Default

min\_age      Minimum age

Type: non-negative integer

Default: No Default

process\_errors      Process error

Type: constant vector

Default: true

process      Process label

Type: string

Default: No Default

`process_proportion`      Process proportion  
Type: constant  
Default: Double(0.5)

`selectivities`      Selectivity labels to use  
Type: string vector  
Default: true

`simulation_likelihood`      Simulation likelihood to use  
Type: string  
Default: ""

`time_step`      Time step to execute in  
Type: string  
Default: No Default

`tolerance`      Tolerance  
Type: constant  
Default: Double(0.001)

`years`      Year to execute in  
Type: non-negative integer vector  
Default: No Default

#### 10.1.6. `@observation[label].type=time_step_proportions_at_age`

`age_plus`      Use age plus group  
Type: boolean  
Default: true

`ageing_error`      Label of ageing error to use  
Type: string  
Default: ""

`categories`      Category labels to use  
Type: string vector  
Default: true

`delta`      Delta  
Type: constant  
Default: DELTA

`error_value_multiplier`      Error value multiplier for likelihood  
Type: constant  
Default: Double(1.0)

likelihoodmultiplier      Likelihood score multiplier

Type: constant

Default: Double(1.0)

likelihood      Type of likelihood to use

Type: string

Default: No Default

max\_age      Maximum age

Type: non-negative integer

Default: No Default

min\_age      Minimum age

Type: non-negative integer

Default: No Default

process\_errors      Process error

Type: constant vector

Default: true

selectivities      Selectivity labels to use

Type: string vector

Default: true

simulation\_likelihood      Simulation likelihood to use

Type: string

Default: ""

time\_step      Time step to execute in

Type: string

Default: No Default

time\_step\_proportion      Proportion through the time step to analyse the partition from

Type: constant

Default: Double(0.5)

tolerance      Tolerance

Type: constant

Default: Double(0.001)

years      Year to execute in

Type: non-negative integer vector

Default: No Default

**10.1.7. @observation[label].type=proportions\_at\_age\_for\_fishery**

age\_plus      Use age plus group

Type: boolean

Default: true

ageing\_error      Label of ageing error to use

Type: string

Default: ""

categories      Category labels to use

Type: string vector

Default: true

delta      Delta

Type: constant

Default: DELTA

error\_value\_multiplier      Error value multiplier for likelihood

Type: constant

Default: Double(1.0)

fishery      Label of fishery the observation is from

Type: string vector

Default: ""

likelihood\_multiplier      Likelihood score multiplier

Type: constant

Default: Double(1.0)

likelihood      Type of likelihood to use

Type: string

Default: No Default

max\_age      Maximum age

Type: non-negative integer

Default: No Default

min\_age      Minimum age

Type: non-negative integer

Default: No Default

process\_errors      Process error

Type: constant vector

Default: true

`process`      **Process label**

Type: string

Default: No Default

`simulation_likelihood`      **Simulation likelihood to use**

Type: string

Default: ""

`time_step`      **Time steps that the fisheries are in**

Type: string vector

Default: No Default

`tolerance`      **Tolerance**

Type: constant

Default: Double(0.001)

`years`      **Year to execute in**

Type: non-negative integer vector

Default: No Default

### **10.1.8. `@observation[label].type=process_proportions_at_length`**

`categories`      **Category labels to use**

Type: string vector

Default: true

`delta`      **Delta**

Type: constant

Default: DELTA

`error_value_multiplier`      **Error value multiplier for likelihood**

Type: constant

Default: Double(1.0)

`length_bins`      **Length bins**

Type: constant vector

Default: No Default

`length_plus_group`      **Is the last bin a plus group**

Type: boolean

Default: true

`likelihood_multiplier`      **Likelihood score multiplier**

Type: constant

Default: Double(1.0)



`likelihood`      Type of likelihood to use  
Type: string  
Default: No Default

`process_errors`      Process error  
Type: constant vector  
Default: true

`process`      Process label  
Type: string  
Default: No Default

`process_proportion`      Process proportion  
Type: constant  
Default: Double(0.5)

`selectivities`      Selectivity labels to use  
Type: string vector  
Default: true

`simulation_likelihood`      Simulation likelihood to use  
Type: string  
Default: ""

`time_step`      Time step to execute in  
Type: string  
Default: No Default

`tolerance`      Tolerance for rescaling proportions  
Type: constant  
Default: Double(0.001)

`years`      Year to execute in  
Type: non-negative integer vector  
Default: No Default

#### 10.1.9. `@observation[label].type=time_step_proportions_at_length`

`categories`      Category labels to use  
Type: string vector  
Default: true

`delta`      Delta

Type: constant  
Default: DELTA

`error_value_multiplier`      Error value multiplier for likelihood  
Type: constant  
Default: Double(1.0)

`length_bins`      Length bins  
Type: constant vector  
Default: No Default

`length_plus_group`      Is the last bin a plus group  
Type: boolean  
Default: true

`likelihood_multiplier`      Likelihood score multiplier  
Type: constant  
Default: Double(1.0)

`likelihood`      Type of likelihood to use  
Type: string  
Default: No Default

`process_errors`      Process error  
Type: constant vector  
Default: true

`selectivities`      Selectivity labels to use  
Type: string vector  
Default: true

`simulation_likelihood`      Simulation likelihood to use  
Type: string  
Default: ""

`time_step`      Time step to execute in  
Type: string  
Default: No Default

`time_step_proportion`      Proportion through the time step to analyse the partition from  
Type: constant  
Default: Double(0.5)

`tolerance`      Tolerance for rescaling proportions

Type: constant  
Default: Double(0.001)

years      Year to execute in  
Type: non-negative integer vector  
Default: No Default

#### 10.1.10. @observation[label].type=proportions\_at\_length\_for\_fishery

categories      Category labels to use  
Type: string vector  
Default: true

delta      Delta  
Type: constant  
Default: DELTA

error\_value\_multiplier      Error value multiplier for likelihood  
Type: constant  
Default: Double(1.0)

fishery      Label of fishery the observation is from  
Type: string  
Default: ""

length\_bins      Length bins  
Type: constant vector  
Default: No Default

length\_plus\_group      Is the last bin a plus group  
Type: boolean  
Default: true

likelihoodmultiplier      Likelihood score multiplier  
Type: constant  
Default: Double(1.0)

likelihood      Type of likelihood to use  
Type: string  
Default: No Default

process\_errors      Process error  
Type: constant vector  
Default: true

`process`      **Process label**

Type: string

Default: No Default

`process_proportion`      **Process proportion**

Type: constant

Default: Double(0.5)

`simulation_likelihood`      **Simulation likelihood to use**

Type: string

Default: ""

`time_step`      **Time step to execute in**

Type: string

Default: No Default

`tolerance`      **Tolerance for rescaling proportions**

Type: constant

Default: Double(0.001)

`years`      **Year to execute in**

Type: non-negative integer vector

Default: No Default

#### **10.1.11. `@observation[label].type=process_proportions_by_category`**

`age_plus`      **Use age plus group**

Type: boolean

Default: true

`categories`      **Category labels to use**

Type: string vector

Default: true

`delta`      **Delta**

Type: constant

Default: DELTA

Lower Bound: 0.0 (exclusive)

`error_value_multiplier`      **Error value multiplier for likelihood**

Type: constant

Default: Double(1.0)

`likelihood_multiplier`      **Likelihood score multiplier**

	Type: constant Default: Double(1.0)
likelihood	Type of likelihood to use Type: string Default: No Default
max_age	Maximum age Type: non-negative integer Default: No Default
min_age	Minimum age Type: non-negative integer Default: No Default
process_errors	Process error Type: constant vector Default: true
process	Process label Type: string Default: No Default
process_proportion	Process proportion Type: constant Default: Double(0.5)
selectivities	Selectivity labels to use Type: string vector Default: true
simulation_likelihood	Simulation likelihood to use Type: string Default: ""
categories2	Target Categories Type: string vector Default: No Default
selectivities2	Target Selectivities Type: string vector Default: No Default
time_step	Time step to execute in

Type: string  
Default: No Default

years      Year to execute in  
Type: non-negative integer vector  
Default: No Default

### 10.1.12. @observation[label].type=time\_step\_proportions\_by\_category

age\_plus      Use age plus group  
Type: boolean  
Default: true

categories      Category labels to use  
Type: string vector  
Default: true

delta      Delta  
Type: constant  
Default: DELTA  
Lower Bound: 0.0 (exclusive)

error\_value\_multiplier      Error value multiplier for likelihood  
Type: constant  
Default: Double(1.0)

likelihood\_multiplier      Likelihood score multiplier  
Type: constant  
Default: Double(1.0)

likelihood      Type of likelihood to use  
Type: string  
Default: No Default

max\_age      Maximum age  
Type: non-negative integer  
Default: No Default

min\_age      Minimum age  
Type: non-negative integer  
Default: No Default

process\_errors      Process error  
Type: constant vector  
Default: true

`selectivities`      Selectivity labels to use

Type: string vector

Default: true

`simulation_likelihood`      Simulation likelihood to use

Type: string

Default: ""

`categories2`      Target Categories

Type: string vector

Default: No Default

`selectivities2`      Target Selectivities

Type: string vector

Default: No Default

`time_step`      Time step to execute in

Type: string

Default: No Default

`time_step_proportion`      Proportion through the time step to analyse the partition from

Type: constant

Default: Double(0.5)

`years`      Year to execute in

Type: non-negative integer vector

Default: No Default

### 10.1.13. `@observation[label].type=proportions_migrating`

`age_plus`      Use age plus group

Type: boolean

Default: true

`ageing_error`      Label of ageing error to use

Type: string

Default: ""

`categories`      Category labels to use

Type: string vector

Default: true

`delta`      Delta

Type: constant  
Default: DELTA

`error_value_multiplier`      Error value multiplier for likelihood  
Type: constant  
Default: Double(1.0)

`likelihood_multiplier`      Likelihood score multiplier  
Type: constant  
Default: Double(1.0)

`likelihood`      Type of likelihood to use  
Type: string  
Default: No Default

`max_age`      Maximum age  
Type: non-negative integer  
Default: No Default

`min_age`      Minimum age  
Type: non-negative integer  
Default: No Default

`process_errors`      Process error  
Type: constant vector  
Default: true

`process`      Process label  
Type: string  
Default: No Default

`process_proportion`      Process proportion  
Type: constant  
Default: Double(0.5)

`simulation_likelihood`      Simulation likelihood to use  
Type: string  
Default: ""

`time_step`      Time step to execute in  
Type: string  
Default: No Default

`years`      Year to execute in



Type: non-negative integer vector

Default: No Default

#### 10.1.14. `@observation[label].type=tag_recapture_by_age`

`age_plus`      Use age plus group

Type: boolean

Default: true

`categories`      Category labels to use

Type: string vector

Default: true

`delta`      Delta

Type: constant

Default: DELTA

Lower Bound: 0.0 (exclusive)

`detection`      Detection probability

Type: constant

Default: No Default

`error_value_multiplier`      Error value multiplier for likelihood

Type: constant

Default: Double(1.0)

`likelihood_multiplier`      Likelihood score multiplier

Type: constant

Default: Double(1.0)

`likelihood`      Type of likelihood to use

Type: string

Default: No Default

`max_age`      Maximum age

Type: non-negative integer

Default: No Default

`min_age`      Minimum age

Type: non-negative integer

Default: No Default

`process_errors`      Process error

Type: constant vector

Default: true

`selectivities`      Selectivity labels to use

Type: string vector

Default: true

`simulation_likelihood`      Simulation likelihood to use

Type: string

Default: ""

`categories2`      Target Categories

Type: string vector

Default: No Default

`selectivities2`      Target Selectivities

Type: string vector

Default: No Default

`time_step`      Time step to execute in

Type: string

Default: No Default

`time_step_proportion`      Proportion through the time step to analyse the partition from

Type: constant

Default: Double(0.5)

`years`      Year to execute in

Type: non-negative integer vector

Default: No Default

### 10.1.15. `@observation[label].type=tag_recapture_by_length`

`categories`      Category labels to use

Type: string vector

Default: true

`delta`      Delta

Type: constant

Default: DELTA

Lower Bound: 0.0 (exclusive)

`detection`      Detection probability

Type: constant

Default: No Default

`error_value_multiplier`      Error value multiplier for likelihood  
Type: constant  
Default: Double(1.0)

`length_bins`      Length Bins  
Type: constant vector  
Default: No Default

`likelihood_multiplier`      Likelihood score multiplier  
Type: constant  
Default: Double(1.0)

`likelihood`      Type of likelihood to use  
Type: string  
Default: No Default

`plus_group`      Last length bin a plus group  
Type: boolean  
Default: true

`process_errors`      Process error  
Type: constant vector  
Default: true

`selectivities`      Selectivity labels to use  
Type: string vector  
Default: true

`simulation_likelihood`      Simulation likelihood to use  
Type: string  
Default: ""

`categories2`      Target Categories  
Type: string vector  
Default: No Default

`selectivities2`      Target Selectivities  
Type: string vector  
Default: No Default

`time_step`      Time step to execute in  
Type: string  
Default: No Default

`time_step_proportion`      Proportion through the time step to analyse the partition from

Type: constant

Default: Double(0.5)

years      Year to execute in

Type: non-negative integer vector

Default: No Default

## 10.2. Likelihoods

**@likelihood** *label*      Define an object type Likelihood

*label*

Type: string

Default: No Default

*type*

Type: string

Default: No Default

**10.2.1. @likelihood[label].type=binomial**

**10.2.2. @likelihood[label].type=binomial\_approx**

**10.2.3. @likelihood[label].type=dirichlet**

**10.2.4. @likelihood[label].type=log\_normal**

**10.2.5. @likelihood[label].type=log\_normal\_with\_q**

**10.2.6. @likelihood[label].type=multinomial**

**10.2.7. @likelihood[label].type=normal**

**10.2.8. @likelihood[label].type=pseudo**

## 10.3. Defining ageing error

Three methods for including ageing error into estimation with observations are,

- None
- Normal
- Off-by-one

Each type of ageing error requires a set of subcommands and arguments specific to its type.

**@ageing.error** *label*      Define an object type Ageing\_Error

label     **Label**  
Type: string  
Default: No Default

type     **Type**  
Type: string  
Default: No Default

### **10.3.1. @ageing\_\_error[label].type=data**

### **10.3.2. @ageing\_\_error[label].type=normal**

cv     **CV for Misclassification matrix**  
Type: constant  
Default: No Default  
Lower Bound: 0.0 (inclusive)

k     **TBA**  
Type: non-negative integer  
Default: 0u

### **10.3.3. @ageing\_\_error[label].type=off\_by\_one**

k     **The minimum age of fish which can be missclassified**  
Type: non-negative integer  
Default: 0u  
Lower Bound: 0.0 (inclusive)  
Upper Bound: 1.0 (inclusive)

p1     **proprtion of misclassification up by an age, i.e. how proportion of age 3 that are actually age 4**  
Type: constant  
Default: No Default

p2     **proprtion of misclassification down by an age**  
Type: constant  
Default: No Default  
Lower Bound: 0.0 (inclusive)  
Upper Bound: 1.0 (inclusive)

## 11. Report command and subcommand syntax

### 11.1. Report commands and subcommands

**@report** *label*     Define an object type Report

*file\_name*     File Name

Type: string

Default: ""

*label*     Label

Type: string

Default: No Default

*type*     Type

Type: string

Default: No Default

*write\_mode*     Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

#### 11.1.1. @report [*label*] .type=ageing\_error\_matrix

*ageing\_error*     Ageing Error label

Type: string

Default: No Default

*file\_name*     File Name

Type: string

Default: ""

*write\_mode*     Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

#### 11.1.2. @report [*label*] .type=category\_info

*file\_name*     File Name

Type: string

Default: ""

*write\_mode*     Write mode

Type: string  
Default: overwrite  
Allowed Values: overwrite, append, incremental\_suffix

### 11.1.3. @report[label].type=category\_list

file\_name      File Name  
Type: string  
Default: ""

write\_mode      Write mode  
Type: string  
Default: overwrite  
Allowed Values: overwrite, append, incremental\_suffix

### 11.1.4. @report[label].type=covariance\_matrix

file\_name      File Name  
Type: string  
Default: ""

write\_mode      Write mode  
Type: string  
Default: overwrite  
Allowed Values: overwrite, append, incremental\_suffix

### 11.1.5. @report[label].type=derived\_quantity

file\_name      File Name  
Type: string  
Default: ""

units      Unit of weight output expressed in  
Type: string  
Default: No Default

write\_mode      Write mode  
Type: string  
Default: overwrite  
Allowed Values: overwrite, append, incremental\_suffix

### 11.1.6. @report[label].type=estimable

file\_name      File Name

Type: string

Default: ""

parameter      Parameter to print

Type: string

Default: No Default

time\_step      Time Step label

Type: string

Default: ""

write\_mode      Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

years      Years to print the estimable for

Type: non-negative integer vector

Default: No Default

### 11.1.7. @report[label].type=estimate\_summary

file\_name      File Name

Type: string

Default: ""

write\_mode      Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

### 11.1.8. @report[label].type=estimate\_value

file\_name      File Name

Type: string

Default: ""

write\_mode      Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix



**11.1.9. @report[label].type=initialisation\_partition**

file\_name      File Name

Type: string

Default: ""

write\_mode      Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

**11.1.10. @report[label].type=mcmc\_covariance**

file\_name      File Name

Type: string

Default: ""

write\_mode      Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

**11.1.11. @report[label].type=mcmc\_objective**

file\_name      File Name

Type: string

Default: ""

write\_mode      Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

**11.1.12. @report[label].type=mcmc\_sample**

file\_name      File Name

Type: string

Default: ""

write\_mode      Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

### 11.1.13. @report [label] .type=m\_p\_d

file\_name      File Name

Type: string

Default: ""

write\_mode      Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

### 11.1.14. @report [label] .type=objective\_function

file\_name      File Name

Type: string

Default: ""

write\_mode      Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

### 11.1.15. @report [label] .type=observation

file\_name      File Name

Type: string

Default: ""

observation      Observation label

Type: string

Default: No Default

write\_mode      Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

### 11.1.16. @report [label] .type=output\_parameters

file\_name      File Name

Type: string

Default: ""

write\_mode      Write mode  
Type: string  
Default: overwrite  
Allowed Values: overwrite, append, incremental\_suffix

#### 11.1.17. @report[label].type=partition

file\_name      File Name  
Type: string  
Default: ""

time\_step      Time Step label  
Type: string  
Default: ""

write\_mode      Write mode  
Type: string  
Default: overwrite  
Allowed Values: overwrite, append, incremental\_suffix

years          Years  
Type: non-negative integer vector  
Default: true

#### 11.1.18. @report[label].type=partition\_biomass

file\_name      File Name  
Type: string  
Default: ""

time\_step      Time Step label  
Type: string  
Default: ""

units          Units (Default Kgs)  
Type: string  
Default: kgs

write\_mode      Write mode  
Type: string  
Default: overwrite  
Allowed Values: overwrite, append, incremental\_suffix

years          Years

Type: non-negative integer vector

Default: true

#### 11.1.19. **@report [label] .type=partition\_mean\_weight**

file\_name      File Name

Type: string

Default: ""

time\_step      Time Step label

Type: string

Default: ""

write\_mode      Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

years      Years

Type: non-negative integer vector

Default: true

#### 11.1.20. **@report [label] .type=process**

file\_name      File Name

Type: string

Default: ""

process      Process label that is reported

Type: string

Default: ""

write\_mode      Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

#### 11.1.21. **@report [label] .type=random\_number\_seed**

file\_name      File Name

Type: string

Default: ""

`write_mode`      Write mode  
Type: string  
Default: overwrite  
Allowed Values: overwrite, append, incremental\_suffix

#### 11.1.22. **@report [label] .type=selectivity**

`file_name`      File Name  
Type: string  
Default: ""

`selectivity`      Selectivity name  
Type: string  
Default: No Default

`write_mode`      Write mode  
Type: string  
Default: overwrite  
Allowed Values: overwrite, append, incremental\_suffix

#### 11.1.23. **@report [label] .type=simulated\_observation**

`file_name`      File Name  
Type: string  
Default: ""

`observation`      Observation label  
Type: string  
Default: No Default

`write_mode`      Write mode  
Type: string  
Default: overwrite  
Allowed Values: overwrite, append, incremental\_suffix

#### 11.1.24. **@report [label] .type=standard\_header**

`file_name`      File Name  
Type: string  
Default: ""

`write_mode`      Write mode

Type: string

Default: overwrite

Allowed Values: overwrite, append, incremental\_suffix

## 12. Other commands and subcommands

**@include** *file*    Include an external file

*file*    The name of the external file to include

Type: string

Default: No default

Value: A valid external file

Condition: The file name must be enclosed in double quotes

Example: @include "my\_file.txt"

Note: @include does not denote the end of the previous command block as is the case for all other commands

---

## 13. Examples

### 13.1. An example of a simple model

This example implements a very simple single species and area model, with recruitment, maturation, natural and fishing mortality, and an annual age increment. The population structure has ages 1 – 30<sup>+</sup> with a single category.

CASAL<sup>2</sup> default file to search for in your current working directory is `casal2.txt`. In this example, `casal2.txt` specifies all the files necessary to run your CASAL<sup>2</sup> model from your current working directory. This is done using the `!include` command as follows.

```
!include "population.csl2"
!include "reports.csl2"
!include "Observation.csl2"
!include "estimation.csl2"
```

Breaking up a CASAL<sup>2</sup> model into sections is recommended, as it aids in readability and error checking. `population.csl2` contains the population information. The model runs from 1975-2012 and is initialised over a 120 year period prior to 1975, which applies the following processes,

1. A Beverton-Holt recruitment process, recruiting a constant number of individuals to the first age class (i.e.,  $age = 1$ ).
2. A constant mortality process representing natural mortality( $M$ ). This process is repeated in all three time steps, so that each with its own time step proportion of  $M$  applied.
3. An ageing process, where all individuals are aged by one year, and with a plus group accumulator age class at  $age = 30$ .

Following initialisation, the model runs from the years 1975 to 2012 iterating through two time-steps. The first time-step applies processes of recruitment, and  $\frac{1}{2}M_1 + F + \frac{1}{2}M_1$  processes, where  $M_1$  is the proportion of  $M$  applied in the first time step. The exploitation process (fishing) is applied in the years 1975–2012. Catches are defined in the catches table and attribute information on each fishery such as selectivity and time-step they are implemented are in the fisheries table in the `@process` block.

The second time-step applies an age increment and the remaining natural mortality.

The first 28 lines of the main section of the `population.csl2` are,

```
## Model Block
@model
start_year 1975
final_year 2012
min_age 1
max_age 30
age_plus true
initialisation_phases iphasel
time_steps step1 step2

## Category Block
@categories
format
names stock
age_lengths age_size

## Initialisation block
@initialisation_phase iphasel
type iterative
years 120

## Annual Cycle definition
@time_step step1
processes Recruitment instant_mort

@time_step step2
processes Ageing instant_mort
```

To carry out a run of the model (to verify that the model runs without any syntax errors), use the command `casal2 -r`. Note that as CASAL<sup>2</sup> looks for a file named `casal2.txt` by default, we can override this. Hypothetically speaking if our model was all written in `Mymodel.txt` we could call it using the `-c` command like `casal2 -r -c Mymodel.txt`.

To run an estimation, and hence estimate the parameters defined in the file `estimation.csl2` (the catchability constant  $q$ , recruitment  $R_0$ , and the selectivity parameters  $a_{50}$  and  $a_{1095}$ ), use `casal2 -e`. Here, we have piped the output to `estimate.log` using the command `casal2 -e > estimate.log`, reports the user defined reports `reports.csl2` from the final iteration of the estimation, and successful convergence printed to screen,

```
Total elapsed time: 1 second
Completed
```

The main part of the output from the estimation run is summarised in the file `estimate.log`, and the final MPD parameter values can be piped out as a separate report, in this case named `paramaters.out`, using the command `casal2 -e -o paramaters.out > estimate.log`.

A profile on the  $R_0$  parameter can also be run, using `casal2 -p > profile.log`. See the examples folder for an example of the output.

KATH note below, will be useful to copy that document across. Examples on Input file specification go to the file `Input File Specification.odt` found in `CASAL2/Documentation/Software Development`

### 13.2. In line declaration

In line declarations can help shorten models by passing @ blocks, for example

```
@observation chatCPUE
type biomass
```



```
catchability [q=6.52606e-005]
time_step one
categories male+female
selectivities chatFselMale chatFselFemale
likelihood lognormal
years 1992:2001
time_step_proportion 1.0
obs 1.50 1.10 0.93 1.33 1.53 0.90 0.68 0.75 0.57 1.23
error_value 0.35
```

```
@estimate
parameter catchability[chatTANbiomass.one].q
type uniform_log
lower_bound 1e-2
upper_bound 1
In line declaration tips
```

In the above code we are defining and estimating catchability without explicitly creating an @catchability block

When you do an inline declaration the new object will be created with the name of the creator's label.<index> where index will be the word if it's one-nine and the number if it's 10+, for example

```
@mortality halfm
selectivities [type=constant; c=1]
```

```
would create
@selectivity halfm.one
```

if there were 10 categories all with there own selectivity the 10<sup>th</sup> selectivity would be labelled;

```
@selectivity halfm.10
```



---

## 14. Post processing output using R

In the downloaded bundle is a R-package that reads CASAL<sup>2</sup> output into R. The CASAL<sup>2</sup> package has only one function `extract()`, which will read in the entire file. The reporting framework is set up so the each `@report` will start with `*` and end with `*end`. If this is not the case the `extract()` function will most likely fail. A post processing package is being developed to then create plots and process the raw input from the `extract()` function.



---

## 15. Troubleshooting

### 15.1. Introduction

### 15.2. Reporting errors

When reporting a bug or problem to the CASAL<sup>2</sup> development team at [casal2@niwa.co.nz](mailto:casal2@niwa.co.nz), please address the following points.

### 15.3. Guidelines for reporting a problem with CASAL<sup>2</sup>

1. Detail the version of CASAL<sup>2</sup> are you using? e.g., “CASAL<sup>2</sup> 2016-05-18 (rev. a03d5dd)Microsoft Windows executable”
2. What operating system or environment are you using? e.g., “IBM-PC Intel CPU running Microsoft Windows 8.1 Enterprise, Service Pack 1”.
3. Give a brief one-line description of the problem, e.g., “a segmentation fault was reported”.
4. If the problem is reproducible, please list the exact steps required to cause it, remembering to include the relevant CASAL<sup>2</sup> configuration file, other input files, and any out generated. Specify the *exact* command line arguments that were used, e.g., “Using the command `***. -* -*` reports a segmentation fault. The input configuration files are attached.”
5. If the problem is not reproducible (only happened once, or occasionally for no apparent reason), please describe the circumstances in which it occurred and the symptoms observed (but note it is much harder to reproduce and hence fix non-reproducible bugs, but if several reports are made over time that relate to the same thing, then this may help to track down the problem), e.g., “CASAL<sup>2</sup> crashed, but I cannot reproduce how I did it. It seemed to be related to a local network crash but I cannot be sure.”
6. If the problem causes any error messages to appear, please give the *exact* text displayed, e.g., `segmentation fault (core dumped)`.
7. Remember to attach all relevant input and output files so that the problem can be reproduced (it can helpful to compress these into a single file). Without these, it is usually not possible to determine the cause of the problem, and we are unlikely to provide any assistance. Note that it is helpful to be as specific as possible when describing the problem.



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## 16. Acknowledgements





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## 17. Quick reference

**@ageing\_error** *label*    Define an object type Ageing\_Error

*label*    Label

*type*    Type

**@ageing\_error[*label*].type=data**

**@ageing\_error[*label*].type=normal**

*cv*    CV for Misclassification matrix

*k*    TBA

**@ageing\_error[*label*].type=off\_by\_one**

*k*    The minimum age of fish which can be missclassified

*p1*    proprtion of misclassification up by an age, i.e. how proportion of age 3 that are actually age 4

*p2*    proprtion of misclassification down by an age

**@age\_length** *label*    Define an object type Age\_Length

*casal\_switch*    A switch to use CASAL Cumulative normal function, note CASAL2 uses the recent BOOST function which differs from the previous CASAL algorithm

*cv\_first*    CV for the first age class

*cv\_last*    CV for last age class

*distribution*    TBA

*label*    Label

*time\_step\_proportions*    the proportion increase of age through the in each time step that corresponds to a length and thus weight increase

*type*    Type

**@age\_length[*label*].type=data**

*by\_length*    Specifies if the linear interpolation of CV's is a linear function of mean length at age. Default is just by age

*casal\_switch*    A switch to use CASAL Cumulative normal function, note CASAL2 uses the recent BOOST function which differs from the previous CASAL algorithm

*cv\_first*    CV for the first age class

*cv\_last*    CV for last age class

*distribution*    TBA

*external\_gaps*

*internal\_gaps*

*length\_weight*    TBA

*time\_step\_proportions*    the proportion increase of age through the in each time step that corresponds to a length and thus weight increase

**@age\_length[*label*].type=none**

*casal\_switch*    A switch to use CASAL Cumulative normal function, note CASAL2 uses the

recent BOOST function which differs from the previous CASAL algorithm

cv\_first CV for the first age class

cv\_last CV for last age class

distribution TBA

time\_step\_proportions the proportion increase of age through the in each time step that corresponds to a length and thus weight increase

**@age\_\_length[label].type=schnute**

a TBA

b TBA

by\_length TBA

casal\_switch A switch to use CASAL Cumulative normal function, note CASAL2 uses the recent BOOST function which differs from the previous CASAL algorithm

cv\_first CV for the first age class

cv\_last CV for last age class

distribution TBA

length\_weight TBA

tau1 TBA

tau2 TBA

time\_step\_proportions the proportion increase of age through the in each time step that corresponds to a length and thus weight increase

y1 TBA

y2 TBA

**@age\_\_length[label].type=von\_bertalanffy**

by\_length Specifies if the linear interpolation of CV's is a linear function of mean length at age. Default is just by age

casal\_switch A switch to use CASAL Cumulative normal function, note CASAL2 uses the recent BOOST function which differs from the previous CASAL algorithm

cv\_first CV for the first age class

cv\_last CV for last age class

distribution TBA

k TBA

length\_weight TBA

linf TBA

t0 TBA

time\_step\_proportions the proportion increase of age through the in each time step that corresponds to a length and thus weight increase

**@catchability label** Define an object type Catchability

label Label

type

**@catchability[label].type=free**

q The catchability amount

**@derived\_quantity label** Define an object type Derived\_Quantity

categories The list of categories to use when calculating the derived quantity

---

label     **Label**  
time\_step\_proportion\_method  
selectivities     The list of selectivities to use when calculating the derived quantity. 1 per category  
time\_step     The time step to calculate the derived quantity after  
time\_step\_proportion  
type     **Type**

**@derived\_\_quantity[label].type=abundance**

categories     The list of categories to use when calculating the derived quantity  
time\_step\_proportion\_method  
selectivities     The list of selectivities to use when calculating the derived quantity. 1 per category  
time\_step     The time step to calculate the derived quantity after  
time\_step\_proportion

**@derived\_\_quantity[label].type=biomass**

categories     The list of categories to use when calculating the derived quantity  
time\_step\_proportion\_method  
selectivities     The list of selectivities to use when calculating the derived quantity. 1 per category  
time\_step     The time step to calculate the derived quantity after  
time\_step\_proportion

**@estimate label**     Define an object type Estimate

estimation\_phase     **TBA**  
label     **Label**  
lower\_bound     The lowest value the parameter is allowed to have  
mcmc     **TBA**  
parameter     The name of the variable to estimate in the model  
prior     The name of the prior to use for the parameter  
same     A list of parameters that are bound to the value of this estimate  
type     **Type**  
upper\_bound     The highest value the parameter is allowed to have

**@estimate[label].type=beta**

a     **A**  
b     **B**  
estimation\_phase     **TBA**  
lower\_bound     The lowest value the parameter is allowed to have  
mcmc     **TBA**  
mu     **Mu**  
parameter     The name of the variable to estimate in the model  
prior     The name of the prior to use for the parameter  
same     A list of parameters that are bound to the value of this estimate  
sigma     **Sigma**  
upper\_bound     The highest value the parameter is allowed to have

**@estimate[label].type=lognormal**

cv      Cv  
estimation\_phase      TBA  
lower\_bound      The lowest value the parameter is allowed to have  
mcmc      TBA  
mu      Mu  
parameter      The name of the variable to estimate in the model  
prior      The name of the prior to use for the parameter  
same      A list of parameters that are bound to the value of this estimate  
upper\_bound      The highest value the parameter is allowed to have

**@estimate[label].type=normal**

cv      Cv  
estimation\_phase      TBA  
lower\_bound      The lowest value the parameter is allowed to have  
mcmc      TBA  
mu      Mu  
parameter      The name of the variable to estimate in the model  
prior      The name of the prior to use for the parameter  
same      A list of parameters that are bound to the value of this estimate  
upper\_bound      The highest value the parameter is allowed to have

**@estimate[label].type=normal\_by\_stdev**

estimation\_phase      TBA  
lower\_bound      The lowest value the parameter is allowed to have  
mcmc      TBA  
mu      Mu  
parameter      The name of the variable to estimate in the model  
prior      The name of the prior to use for the parameter  
same      A list of parameters that are bound to the value of this estimate  
sigma      Sigma  
upper\_bound      The highest value the parameter is allowed to have

**@estimate[label].type=normal\_log**

estimation\_phase      TBA  
lower\_bound      The lowest value the parameter is allowed to have  
mcmc      TBA  
mu      Mu  
parameter      The name of the variable to estimate in the model  
prior      The name of the prior to use for the parameter  
same      A list of parameters that are bound to the value of this estimate  
sigma      Sigma  
upper\_bound      The highest value the parameter is allowed to have

**@estimate[label].type=uniform**

estimation\_phase      TBA

---

lower\_bound     The lowest value the parameter is allowed to have  
mcmc     TBA  
parameter     The name of the variable to estimate in the model  
prior     The name of the prior to use for the parameter  
same     A list of parameters that are bound to the value of this estimate  
upper\_bound     The highest value the parameter is allowed to have

**@estimate[label].type=uniform\_log**

estimation\_phase     TBA  
lower\_bound     The lowest value the parameter is allowed to have  
mcmc     TBA  
parameter     The name of the variable to estimate in the model  
prior     The name of the prior to use for the parameter  
same     A list of parameters that are bound to the value of this estimate  
upper\_bound     The highest value the parameter is allowed to have  
**@initialisation\_phase label**     Define an object type Initialisation.Phase  
label     Label  
type     Type

**@initialisation\_phase[label].type=cinitial**

categories     List of categories to use

**@initialisation\_phase[label].type=derived**

casal\_initialisation\_switch     Reset the partition after running an extra annual cycle to take on equilibrium SSB's. Warning should only be set to true if comparing with previous CASAL models  
exclude\_processes     The processes to exclude from all time steps  
insert\_processes     The processes to insert in to target time steps

**@initialisation\_phase[label].type=iterative**

convergence\_years     The years to test for convergence  
exclude\_processes     The processes to exclude from all time steps  
insert\_processes     The processes to insert in to target time steps  
lambda     Lambda  
years     The number of iterations to execute this phase for

**@initialisation\_phase[label].type=state\_category\_by\_age**

categories     List of categories to use  
max\_age     Maximum age to use for this process  
min\_age     Minimum age to use for this process  
**@likelihood label**     Define an object type Likelihood  
label  
type

**@likelihood[label].type=binomial**

**@likelihood[label].type=binomial\_approx**

**@likelihood[label].type=dirichlet**

**@likelihood[label].type=log\_normal**

**@likelihood[label].type=log\_normal\_with\_q**

**@likelihood[label].type=multinomial**

**@likelihood[label].type=normal**

**@likelihood[label].type=pseudo**

**@derived\_quantity label** Define an object type `Derived_Quantity`

`categories` The list of categories to use when calculating the derived quantity

`label` Label

`time_step_proportion_method`

`selectivities` The list of selectivities to use when calculating the derived quantity. 1 per category

`time_step` The time step to calculate the derived quantity after

`time_step_proportion`

`type` Type

**@derived\_quantity[label].type=abundance**

`categories` The list of categories to use when calculating the derived quantity

`time_step_proportion_method`

`selectivities` The list of selectivities to use when calculating the derived quantity. 1 per category

`time_step` The time step to calculate the derived quantity after

`time_step_proportion`

**@derived\_quantity[label].type=biomass**

`categories` The list of categories to use when calculating the derived quantity

`time_step_proportion_method`

`selectivities` The list of selectivities to use when calculating the derived quantity. 1 per category

`time_step` The time step to calculate the derived quantity after

`time_step_proportion`

**@mcmc label** Define an object type `MCMC`

`active` Is this the active MCMC algorithm

`label` Label

`length` The number of chain links to create

`print_default_reports`

`type` Type

**@mcmc[label].type=independence\_metropolis**

`active` Is this the active MCMC algorithm

`adapt_stepsize_at` Iterations in the chain to check and resize the MCMC stepsize

`correlation_adjustment_diff` TBA

`covariance_adjustment_method` Method for adjusting small variances in the covariance

---

proposal matrix  
df      Degrees of freedom of the multivariate t proposal distribution  
keep     Spacing between recorded values in the chain  
length    The number of chain links to create  
max\_correlation      Maximum absolute correlation in the covariance matrix of the proposal distribution  
print\_default\_reports  
proposal\_distribution      The shape of the proposal distribution (either t or normal  
start      Covariance multiplier for the starting point of the Markov chain  
step\_size      Initial stepsize (as a multiplier of the approximate covariance matrix  
**@minimiser label**      Define an object type Minimiser  
active      True if this minimiser is active  
covariance      True if a covariance matrix should be created  
label      Label  
type      Type of minimiser to use

**@minimiser[label].type=callback\_a\_d\_o\_l\_c**

active      True if this minimiser is active  
covariance      True if a covariance matrix should be created  
tolerance      Tolerance of the gradient for convergence  
evaluations      Maximum number of evaluations  
iterations      Maximum number of iterations  
step\_size      Minimum Step-size before minimisation fails

**@minimiser[label].type=engine\_a\_d\_o\_l\_c**

active      True if this minimiser is active  
covariance      True if a covariance matrix should be created  
tolerance      Tolerance of the gradient for convergence  
evaluations      Maximum number of evaluations  
iterations      Maximum number of iterations  
step\_size      Minimum Step-size before minimisation fails

**@minimiser[label].type=f\_m\_m\_a\_d\_o\_l\_c**

active      True if this minimiser is active  
covariance      True if a covariance matrix should be created  
tolerance      Tolerance of the gradient for convergence  
evaluations      Maximum number of evaluations  
iterations      Maximum number of iterations  
step\_size      Minimum Step-size before minimisation fails

**@minimiser[label].type=beta\_diff**

active      True if this minimiser is active  
covariance      True if a covariance matrix should be created  
tolerance      Tolerance of the gradient for convergence  
evaluations      Maximum number of evaluations  
iterations      Maximum number of iterations

**@minimiser[label].type=c\_p\_p\_a\_d**

active     True if this minimiser is active  
covariance    True if a covariance matrix should be created

**@minimiser[label].type=call\_back\_d\_e\_solver**

active     True if this minimiser is active  
covariance    True if a covariance matrix should be created  
crossover\_probability    TBA  
difference\_scale    The scale to apply to new solutions when comparing candidates  
max\_generations    The maximum number of iterations to run  
method     The type of candidate generation method to use  
population\_size    The number of candidate solutions to have in the population  
tolerance    The total variance between the population and best candidate before acceptance

**@minimiser[label].type=engine\_d\_e\_solver**

active     True if this minimiser is active  
covariance    True if a covariance matrix should be created  
crossover\_probability    TBA  
difference\_scale    The scale to apply to new solutions when comparing candidates  
max\_generations    The maximum number of iterations to run  
method     The type of candidate generation method to use

**@minimiser[label].type=call\_back\_d\_lib**

active     True if this minimiser is active  
covariance    True if a covariance matrix should be created

**@minimiser[label].type=dummy**

active     True if this minimiser is active  
covariance    True if a covariance matrix should be created

**@minimiser[label].type=callback\_gamma\_diff**

active     True if this minimiser is active  
covariance    True if a covariance matrix should be created  
tolerance    Tolerance of the gradient for convergence  
evaluations    Maximum number of evaluations  
iterations    Maximum number of iterations  
step\_size    Minimum Step-size before minimisation fails

**@minimiser[label].type=engine\_gamma\_diff**

active     True if this minimiser is active



---

covariance    True if a covariance matrix should be created  
tolerance    Tolerance of the gradient for convergence  
evaluations    Maximum number of evaluations  
iterations    Maximum number of iterations  
step\_size    Minimum Step-size before minimisation fails

**@minimiser[label].type=f.m.m\_gamma\_diff**

active    True if this minimiser is active  
covariance    True if a covariance matrix should be created  
tolerance    Tolerance of the gradient for convergence  
evaluations    Maximum number of evaluations  
iterations    Maximum number of iterations  
step\_size    Minimum Step-size before minimisation fails

**@model label**    Define an object type Model

age\_plus    Define the oldest age as a plus group  
final\_year    Define the final year of the model, excluding years in the projection period  
initialisation\_phases    Define the labels of the phases of the initialisation  
label  
length\_bins  
max\_age    Maximum age of individuals in the population  
min\_age    Minimum age of individuals in the population  
projection\_final\_year    Define the final year of the model in projection mode  
start\_year    Define the first year of the model, immediately following initialisation  
time\_steps    Define the labels of the time steps, in the order that they are applied, to form the  
annual cycle  
type    Type of model (the partition structure). Either age, length or hybrid

**@observation label**    Define an object type Observation

categories    Category labels to use  
error\_value\_multiplier    Error value multiplier for likelihood  
label    Label  
likelihood\_multiplier    Likelihood score multiplier  
likelihood    Type of likelihood to use  
simulation\_likelihood    Simulation likelihood to use  
type    Type of observation

**@observation[label].type=process\_abundance**

catchability    Abundance catchability

categories    Category labels to use  
delta        Delta value for error values  
error\_value\_multiplier    Error value multiplier for likelihood  
error\_value    The error values to use against the observation values  
likelihood\_multiplier    Likelihood score multiplier  
likelihood    Type of likelihood to use  
obs        Observation values  
process\_error    Process error  
process        Process label  
process\_proportion    Process proportion  
selectivities    Selectivity labels to use  
simulation\_likelihood    Simulation likelihood to use  
time\_step    Time step to execute in  
years        Years to execute in

**@observation[label].type=time\_step\_abundance**

catchability    TBA  
categories    Category labels to use  
delta        Delta value for error values  
error\_value\_multiplier    Error value multiplier for likelihood  
error\_value    The error values to use against the observation values  
likelihood\_multiplier    Likelihood score multiplier  
likelihood    Type of likelihood to use  
obs        Observation values  
process\_error    Process error  
selectivities    Selectivity labels to use  
simulation\_likelihood    Simulation likelihood to use  
time\_step    Time step to execute in  
time\_step\_proportion    Proportion through the time step to analyse the partition from  
years        Years to execute in

**@observation[label].type=process\_biomass**

catchability    Catchability of Biomass  
categories    Category labels to use  
delta        Delta value for error values  
error\_value\_multiplier    Error value multiplier for likelihood  
error\_value    The error values to use against the observation values  
likelihood\_multiplier    Likelihood score multiplier  
likelihood    Type of likelihood to use  
obs        Observation values  
process\_error    Process error  
process        Process label  
process\_proportion    Process proportion  
selectivities    Selectivity labels to use  
simulation\_likelihood    Simulation likelihood to use  
time\_step    Time step to execute in  
years        Years to execute in

---

### **@observation[label].type=time\_step\_biomass**

catchability    Catchability of Biomass  
categories    Category labels to use  
delta    Delta value for error values  
error\_value\_multiplier    Error value multiplier for likelihood  
error\_value    The error values to use against the observation values  
likelihood\_multiplier    Likelihood score multiplier  
likelihood    Type of likelihood to use  
obs    Observation values  
process\_error    Process error  
selectivities    Selectivity labels to use  
simulation\_likelihood    Simulation likelihood to use  
time\_step    Time step to execute in  
time\_step\_proportion    Proportion through the time step to analyse the partition from  
years    Years to execute in

### **@observation[label].type=process\_proportions\_at\_age**

age\_plus    Use age plus group  
ageing\_error    Label of ageing error to use  
categories    Category labels to use  
delta    Delta  
error\_value\_multiplier    Error value multiplier for likelihood  
likelihood\_multiplier    Likelihood score multiplier  
likelihood    Type of likelihood to use  
max\_age    Maximum age  
min\_age    Minimum age  
process\_errors    Process error  
process    Process label  
process\_proportion    Process proportion  
selectivities    Selectivity labels to use  
simulation\_likelihood    Simulation likelihood to use  
time\_step    Time step to execute in  
tolerance    Tolerance  
years    Year to execute in

### **@observation[label].type=time\_step\_proportions\_at\_age**

age\_plus    Use age plus group

ageing\_error      Label of ageing error to use  
categories        Category labels to use  
delta            Delta  
error\_value\_multiplier      Error value multiplier for likelihood  
likelihood\_multiplier      Likelihood score multiplier  
likelihood        Type of likelihood to use  
max\_age          Maximum age  
min\_age          Minimum age  
process\_errors    Process error  
selectivities      Selectivity labels to use  
simulation\_likelihood      Simulation likelihood to use  
time\_step        Time step to execute in  
time\_step\_proportion      Proportion through the time step to analyse the partition from  
tolerance        Tolerance  
years            Year to execute in

**@observation[label].type=proportions\_at\_age\_for\_fishery**

age\_plus          Use age plus group  
ageing\_error      Label of ageing error to use  
categories        Category labels to use  
delta            Delta  
error\_value\_multiplier      Error value multiplier for likelihood  
fishery          Label of fishery the observation is from  
likelihood\_multiplier      Likelihood score multiplier  
likelihood        Type of likelihood to use  
max\_age          Maximum age  
min\_age          Minimum age  
process\_errors    Process error  
process          Process label  
simulation\_likelihood      Simulation likelihood to use  
time\_step        Time steps that the fisheries are in  
tolerance        Tolerance  
years            Year to execute in

**@observation[label].type=process\_proportions\_at\_length**

categories        Category labels to use

---

delta      Delta  
error\_value\_multiplier      Error value multiplier for likelihood  
length\_bins      Length bins  
length\_plus\_group      Is the last bin a plus group  
likelihood\_multiplier      Likelihood score multiplier  
likelihood      Type of likelihood to use  
process\_errors      Process error  
process      Process label  
process\_proportion      Process proportion  
selectivities      Selectivity labels to use  
simulation\_likelihood      Simulation likelihood to use  
time\_step      Time step to execute in  
tolerance      Tolerance for rescaling proportions  
years      Year to execute in

**@observation[label].type=time\_step\_proportions\_at\_length**

categories      Category labels to use  
delta      Delta  
error\_value\_multiplier      Error value multiplier for likelihood  
length\_bins      Length bins  
length\_plus\_group      Is the last bin a plus group  
likelihood\_multiplier      Likelihood score multiplier  
likelihood      Type of likelihood to use  
process\_errors      Process error  
selectivities      Selectivity labels to use  
simulation\_likelihood      Simulation likelihood to use  
time\_step      Time step to execute in  
time\_step\_proportion      Proportion through the time step to analyse the partition from  
tolerance      Tolerance for rescaling proportions  
years      Year to execute in

**@observation[label].type=proportions\_at\_length\_for\_fishery**

categories      Category labels to use  
delta      Delta  
error\_value\_multiplier      Error value multiplier for likelihood  
fishery      Label of fishery the observation is from  
length\_bins      Length bins  
length\_plus\_group      Is the last bin a plus group  
likelihood\_multiplier      Likelihood score multiplier  
likelihood      Type of likelihood to use  
process\_errors      Process error  
process      Process label  
process\_proportion      Process proportion  
simulation\_likelihood      Simulation likelihood to use  
time\_step      Time step to execute in  
tolerance      Tolerance for rescaling proportions  
years      Year to execute in

**@observation[label].type=process\_proportions\_by\_category**

age\_plus      Use age plus group  
categories    Category labels to use  
delta        Delta  
error\_value\_multiplier    Error value multiplier for likelihood  
likelihood\_multiplier    Likelihood score multiplier  
likelihood    Type of likelihood to use  
max\_age      Maximum age  
min\_age      Minimum age  
process\_errors    Process error  
process       Process label  
process\_proportion    Process proportion  
selectivities    Selectivity labels to use  
simulation\_likelihood    Simulation likelihood to use  
categories2    Target Categories  
selectivities2    Target Selectivities  
time\_step    Time step to execute in  
years        Year to execute in

**@observation[label].type=time\_step\_proportions\_by\_category**

age\_plus      Use age plus group  
categories    Category labels to use  
delta        Delta  
error\_value\_multiplier    Error value multiplier for likelihood  
likelihood\_multiplier    Likelihood score multiplier  
likelihood    Type of likelihood to use  
max\_age      Maximum age  
min\_age      Minimum age  
process\_errors    Process error  
selectivities    Selectivity labels to use  
simulation\_likelihood    Simulation likelihood to use  
categories2    Target Categories  
selectivities2    Target Selectivities  
time\_step    Time step to execute in  
time\_step\_proportion    Proportion through the time step to analyse the partition from  
years        Year to execute in

**@observation[label].type=proportions\_migrating**

age\_plus      Use age plus group

---

ageing\_error      Label of ageing error to use  
categories      Category labels to use  
delta      Delta  
error\_value\_multiplier      Error value multiplier for likelihood  
likelihood\_multiplier      Likelihood score multiplier  
likelihood      Type of likelihood to use  
max\_age      Maximum age  
min\_age      Minimum age  
process\_errors      Process error  
process      Process label  
process\_proportion      Process proportion  
simulation\_likelihood      Simulation likelihood to use  
time\_step      Time step to execute in  
years      Year to execute in

**@observation[label].type=tag\_recapture\_by\_age**

age\_plus      Use age plus group  
categories      Category labels to use  
delta      Delta  
detection      Detection probability  
error\_value\_multiplier      Error value multiplier for likelihood  
likelihood\_multiplier      Likelihood score multiplier  
likelihood      Type of likelihood to use  
max\_age      Maximum age  
min\_age      Minimum age  
process\_errors      Process error  
selectivities      Selectivity labels to use  
simulation\_likelihood      Simulation likelihood to use  
categories2      Target Categories  
selectivities2      Target Selectivities  
time\_step      Time step to execute in  
time\_step\_proportion      Proportion through the time step to analyse the partition from  
years      Year to execute in

**@observation[label].type=tag\_recapture\_by\_length**

categories      Category labels to use

delta      **Delta**  
detection      **Detection probability**  
error\_value\_multiplier      **Error value multiplier for likelihood**  
length.bins      **Length Bins**  
likelihood\_multiplier      **Likelihood score multiplier**  
likelihood      **Type of likelihood to use**  
plus\_group      **Last length bin a plus group**  
process\_errors      **Process error**  
selectivities      **Selectivity labels to use**  
simulation\_likelihood      **Simulation likelihood to use**  
categories2      **Target Categories**  
selectivities2      **Target Selectivities**  
time\_step      **Time step to execute in**  
time\_step\_proportion      **Proportion through the time step to analyse the partition from**  
years      **Year to execute in**  
**@penalty** *label*      **Define an object type Penalty**  
  
label      **Label**  
type      **Type**

**@penalty[label].type=process**

log\_scale      **Log scale**  
multiplier      **Multiplier**  
**@process** *label*      **Define an object type Process**  
  
print\_report      **Generate parameter report**  
label      **Label**  
type      **Type**

**@process[label].type=ageing**

categories      **Categories**  
print\_report      **Generate parameter report**

**@process[label].type=growth**

print\_report      **Generate parameter report**

**@process[label].type=maturation**

print\_report      **Generate parameter report**  
from      **List of categories to mature from**  
rates      **The rates to mature for each year**  
selectivities      **List of selectivities to use for maturation**  
to      **List of categories to mature too**  
years      **The years to be associated with rates**

**@process[label].type=mortality\_constant\_rate**

categories      **List of categories**



---

print\_report    Generate parameter report  
m      Mortality rates  
time\_step\_ratio    Time step ratios for M  
selectivities    Selectivities

**@process[label].type=mortality\_event**

catches    Catches  
categories    Categories  
print\_report    Generate parameter report  
penalty    Penalty label  
selectivities    List of selectivities  
u\_max    U Max  
years    Years

**@process[label].type=mortality\_event\_biomass**

catches    Catches for each year  
categories    Category labels  
print\_report    Generate parameter report  
penalty    Penalty label  
selectivities    Selectivity labels  
u\_max    U Max  
units    Unit of weight that the Catches table are expressed in  
years    Years to apply mortality

**@process[label].type=mortality\_holling\_rate**

a    parameter a  
b    parameter b  
print\_report    Generate parameter report  
penalty    Label of penalty to be applied  
predator\_categories    Predator Categories labels  
predator\_selectivities    Selectivities for predator categories  
prey\_categories    Prey Categories labels  
prey\_selectivities    Selectivities for prey categories  
u\_max    Umax  
x    parameter x  
years    Year to execute in

**@process[label].type=mortality\_instantaneous**

categories    Categories for natural mortality  
print\_report    Generate parameter report  
m    Mortality rates  
selectivities    Selectivities for Natural Mortality  
time\_step\_ratio    Time step ratios for M  
units    Unit of weight that the Catches table are expressed in

**@process[label].type=mortality\_prey\_suitability**

consumption\_rate    Predator consumption rate  
print\_report        Generate parameter report  
electivities        Prey Electivities  
penalty            Label of penalty to be applied  
predator\_categories    Predator Categories labels  
predator\_selectivities    Selectivities for predator categories  
prey\_categories        Prey Categories labels  
prey\_selectivities    Selectivities for prey categories  
u\_max            Umax  
years            Year that process occurs

**@process[label].type=nop**

print\_report        Generate parameter report

**@process[label].type=recruitment\_beverton\_holt**

age            Age to recruit at  
b0            B0  
units        Units of B0, if initialising model using B0  
categories    Category labels  
print\_report    Generate parameter report  
b0\_initialisation\_phase    Initialisation phase Label that b0 is from  
prior\_standardised\_ycs    Priors for year class strength on ycs values (not standardised ycs values)  
proportions    Proportions  
r0            R0  
ssb            SSB Label (derived quantity  
ssb\_offset    Spawning biomass year offset  
standardise\_ycs\_years    Years that are included for year class standardisation  
steepness    Steepness  
ycs\_values    YCS Values

**@process[label].type=recruitment\_constant**

age            Age  
categories    Categories  
print\_report    Generate parameter report  
proportions    Proportions  
r0            R0

**@process[label].type=tag\_by\_age**

print\_report        Generate parameter report

---

```

from      Categories to transition from
initial_mortality
initial_mortality_selectivity
loss_rate
loss_rate_selectivities
max_age    Maximum age to transition
min_age    Minimum age to transition
n
penalty    Penalty label
selectivities
to      Categories to transition to
u_max     U Max
years     Years to execute the transition in

```

#### **@process[label].type=tag\_by\_length**

```

print_report    Generate parameter report
from      Categories to transition from
initial_mortality
initial_mortality_selectivity
maximum_length    The upper length when there is no plus group
n
penalty    Penalty label
plus_group    Use plus group for last length bin
selectivities
to      Categories to transition to
u_max     U Max
years     Years to execute the transition in

```

#### **@process[label].type=tag\_loss**

```

categories    List of categories
print_report    Generate parameter report
time_step_ratio    Time step ratios for Tag Loss
selectivities    Selectivities
tag_loss_rate    Tag Loss rates
tag_loss_type    Type of tag loss
year    The year the first tagging release process was executed

```

#### **@process[label].type=transition\_category**

```

print_report    Generate parameter report
from      From
proportions    Proportions
selectivities    Selectivity names
to      To

```

#### **@process[label].type=transition\_category\_by\_age**

```

print_report    Generate parameter report

```

from      Categories to transition from  
max\_age    Maximum age to transition  
min\_age    Minimum age to transition  
penalty    Penalty label  
to        Categories to transition to  
u\_max      U Max  
years      Years to execute the transition in  
**@profile** *label*    Define an object type Profile  
  
label      Label  
lower\_bound    The lower bounds  
parameter    The system parameter to profile  
steps      The number of steps to take between the lower and upper bound  
type  
upper\_bound    The upper bounds  
**@report** *label*    Define an object type Report  
  
file\_name    File Name  
label      Label  
type      Type  
write\_mode    Write mode

**@report [label] .type=ageing\_error\_matrix**

ageing\_error    Ageing Error label  
file\_name      File Name  
write\_mode      Write mode

**@report [label] .type=category\_info**

file\_name    File Name  
write\_mode    Write mode

**@report [label] .type=category\_list**

file\_name    File Name  
write\_mode    Write mode

**@report [label] .type=covariance\_matrix**

file\_name    File Name  
write\_mode    Write mode

**@report [label] .type=derived\_quantity**

file\_name    File Name  
units      Unit of weight output expressed in  
write\_mode    Write mode

**@report [label] .type=estimable**

file\_name    File Name

---

parameter     Parameter to print  
time\_step     Time Step label  
write\_mode     Write mode  
years     Years to print the estimable for

**@report[label].type=estimate\_summary**

file\_name     File Name  
write\_mode     Write mode

**@report[label].type=estimate\_value**

file\_name     File Name  
write\_mode     Write mode

**@report[label].type=initialisation\_partition**

file\_name     File Name  
write\_mode     Write mode

**@report[label].type=mcmc\_covariance**

file\_name     File Name  
write\_mode     Write mode

**@report[label].type=mcmc\_objective**

file\_name     File Name  
write\_mode     Write mode

**@report[label].type=mcmc\_sample**

file\_name     File Name  
write\_mode     Write mode

**@report[label].type=m\_p\_d**

file\_name     File Name  
write\_mode     Write mode

**@report[label].type=objective\_function**

file\_name     File Name  
write\_mode     Write mode

**@report[label].type=observation**

file\_name     File Name  
observation     Observation label  
write\_mode     Write mode

**@report[label].type=output\_parameters**

file\_name     File Name  
write\_mode    Write mode

**@report[label].type=partition**

file\_name     File Name  
time\_step     Time Step label  
write\_mode    Write mode  
years        Years

**@report[label].type=partition\_biomass**

file\_name     File Name  
time\_step     Time Step label  
units        Units (Default Kgs  
write\_mode    Write mode  
years        Years

**@report[label].type=partition\_mean\_weight**

file\_name     File Name  
time\_step     Time Step label  
write\_mode    Write mode  
years        Years

**@report[label].type=process**

file\_name     File Name  
process       Process label that is reported  
write\_mode    Write mode

**@report[label].type=random\_number\_seed**

file\_name     File Name  
write\_mode    Write mode

**@report[label].type=selectivity**

file\_name     File Name  
selectivity    Selectivity name  
write\_mode    Write mode

**@report[label].type=simulated\_observation**

file\_name     File Name  
observation    Observation label  
write\_mode    Write mode

**@report[label].type=standard\_header**

---

file\_name    File Name  
write\_mode    Write mode  
**@selectivity** *label*    Define an object type Selectivity  
label    Label  
length\_based    Is the selectivity length based  
intervals    Number of quantiles to evaluate a length based selectivity over the age length  
distribution  
type    Type

**@selectivity[label].type=all\_values**

length\_based    Is the selectivity length based  
intervals    Number of quantiles to evaluate a length based selectivity over the age length  
distribution  
v    V

**@selectivity[label].type=all\_values\_bounded**

h    H  
length\_based    Is the selectivity length based  
l    L  
intervals    Number of quantiles to evaluate a length based selectivity over the age length  
distribution  
v    V

**@selectivity[label].type=constant**

c    C  
length\_based    Is the selectivity length based  
intervals    Number of quantiles to evaluate a length based selectivity over the age length  
distribution

**@selectivity[label].type=double\_exponential**

alpha    Alpha  
length\_based    Is the selectivity length based  
intervals    Number of quantiles to evaluate a length based selectivity over the age length  
distribution  
x0    X0  
x1    X1  
x2    X2  
y0    Y0  
y1    Y1  
y2    Y2

**@selectivity[label].type=double\_normal**

alpha    Alpha

length\_based    Is the selectivity length based  
mu      Mu  
intervals      Number of quantiles to evaluate a length based selectivity over the age length  
distribution  
sigma\_l      Sigma L  
sigma\_r      Sigma R

**@selectivity[label].type=increasing**

alpha    Alpha  
h      High  
length\_based    Is the selectivity length based  
l      Low  
intervals      Number of quantiles to evaluate a length based selectivity over the age length  
distribution  
v      V

**@selectivity[label].type=inverse\_logistic**

a50    A50  
alpha    Alpha  
ato95    aTo95  
length\_based    Is the selectivity length based  
intervals      Number of quantiles to evaluate a length based selectivity over the age length  
distribution

**@selectivity[label].type=knife\_edge**

alpha    Alpha  
e      Edge  
length\_based    Is the selectivity length based  
intervals      Number of quantiles to evaluate a length based selectivity over the age length  
distribution

**@selectivity[label].type=logistic**

a50    A50  
alpha    Alpha  
ato95    Ato95  
length\_based    Is the selectivity length based  
intervals      Number of quantiles to evaluate a length based selectivity over the age length  
distribution

**@selectivity[label].type=logistic\_producing**

a50    A50



---

```

alpha      Alpha
ato95      Ato95
h          High
length_based  Is the selectivity length based
l          Low
intervals  Number of quantiles to evaluate a length based selectivity over the age length
distribution
@length_weight label  Define an object type Length_Weight
label      Label
type       Type

@length_weight[label].type=basic
a          A
b          B
units      Units of measure (tonnes, kgs, grams)

@length_weight[label].type=None
@time_step label  Define an object type Time_Step
label      Label
processes  Processes
type

```



---

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