# DLMtool: Data-Limited Methods Toolkit (v3.2) 

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# Package 'DLMtool' 

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DLMtool-package Data-Limited Methods Toolkit

## Description

Simulation testing and implementation of data-limited fishery stock assessment methods

## Details

| Package: | DLMtool |
| :--- | :--- |
| Type: | Package |
| Version: | 3.2 |
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| License: | GPL-2 |
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## Author(s)

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## References

Carruthers, T.R., Punt, A.E., Walters, C.J., MacCall, A., McAllister, M.K., Dick, E.J., Cope, J. 2014. Evaluating methods for setting catch limits in data-limited fisheries. Fisheries Research. 153: 48-68.

Carruthers, T.R., Kell, L.T., Butterworth, D.S., Maunder, M.N., Geromont, H.F., Walters, C., McAllister, M.K., Hillary, R., Levontin, P., Kitakado, T., Davies, C.R. Performance review of simple management procedures. ICES Journal of Marine Science.

## Examples

```
# --- Application to real fishery data ---
library(DLMtool)
library(snowfall) # load package for parallel computing
sfInit(parallel=TRUE,cpus=2) # initiate the cluster with two cpus
```

```
mydata<-new('DLM_data') # create a new DLM data object and define:
mydata@Year<-2001:2010 # years
mydata@Cat<-matrix((11:20)*10*runif(10,0.5,1.5),nrow=1) # make up some annual catches
mydata@Ind<-matrix(seq(1.1,0.9,length.out=10)*runif(10,0.5,1.5),nrow=1)
mydata@Mort<-0.2 # instantaneous natural mortality rate
mydata@Abun<-1000 # current abundance estimate (biomass)
mydata@FMSY_M<-0.5 # guess of the ratio of FMSY to natural mortality rate
mydata@vbLinf<-200 # maximum length
mydata@vbK<-0.2 # von B growth coefficient k
mydata@LFC<-50 # length at first capture
mydata<-TAC(mydata) # calculate quotas
plot(mydata) # plot them
mydata<-Sense(mydata,"Fratio") # conduct a sensitivity analysis for one of the methods
sfStop()
```

    avail What objects of this class are available
    
## Description

Generic class finder

## Usage

```
avail(classy)
```


## Arguments

classy A class of object (character string, e.g. 'Fleet')

## Details

Finds objects of the specified class in the global environment or the package:DLMtool

## Author(s)

T. Carruthers

## Description

A simple average catch MP that is included to demonstrate a 'status quo' management option.

## Usage

AvC(x, DLM_data, reps = 100)

## Arguments

x
A position in a data-limited methods data object
DLM_data
A data-limited methods data object
reps
The number of stochastic samples of the quota recommendation

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers

## Description

Sets an OFL according to current abundance and an approximation of FMSY based on length at first capture.

## Usage

$B K\left(x, D L M \_d a t a\right.$, reps $\left.=100\right)$

## Arguments

x
DLM_data
reps

A position in a data-limited methods data object.
A data-limited methods data object.
The number of stochastic samples of the TAC recommendation

## Note

This is the simple version of the BK MP. The paper has a more complex approach that might work better.

## Author(s)

T. Carruthers.

## References

Beddington, J.R., Kirkwood, G.P., 2005. The estimation of potential yield and stock status using life history parameters. Philos. Trans. R. Soc. Lond. B Biol. Sci. 360, 163-170.

```
BK_CC
```

Beddington and Kirkwood life-history method combined with catch curve analysis

## Description

Calculates an OFL using a catch curve estimate of current F and an approximation of FMSY based on length at first capture.

## Usage

BK_CC(x, DLM_data, reps $=100$, Fmin=0.005)

## Arguments

$x \quad$ Position in a data-limited methods data object
DLM_data A data-limited methods data object (class DLM_data)
reps $\quad$ The number of samples of the TAC recommendation
Fmin The minimum fishing mortality rate that is derived from the catch-curve (interval censor)

## Author(s)

T. Carruthers

## References

Beddington, J.R., Kirkwood, G.P., 2005. The estimation of potential yield and stock status using life history parameters. Philos. Trans. R. Soc. Lond. B Biol. Sci. 360, 163-170.
BK_ML Beddington and Kirkwood life-history analysis with mean-length esti- mator of current abundance

## Description

Uses an approximation to FMSY based on length at first capture and an estimate of current abundance based on a mean-length estimator.

## Usage

BK_ML(x, DLM_data, reps = 100)

## Arguments

| x | Position in a data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object (class DLM_data) |
| reps | The number of samples of the TAC recommendation |

## Note

The mean length extension was programmed by Gary Nelson as part of his excellent R package 'fishmethods'

## Author(s)

T. Carruthers

## References

Beddington, J.R., Kirkwood, G.P., 2005. The estimation of potential yield and stock status using life history parameters. Philos. Trans. R. Soc. Lond. B Biol. Sci. 360, 163-170.

## Can

What data-limited methods can be applied to this DLM_data object?

## Description

An diagnostic tool that looks up the slot requirements of each method and compares this to the data available to limit the analysis to methods that have the correct data, do not produce errors and run within a time limit. Time limit is the maximum time taken to carry out five reps (stochastic samples) of a given method and is in units of seconds.

## Usage

Can(DLM_data, timelimit = 1)

## Arguments

DLM_data A data-limited methods data object (class DLM_data)
timelimit The maximum time (seconds) taken for a method to undertake 10 reps (this filters out methods that are too slow)
Cant What methods can't be applied to this DLM data object

## Description

The methods that don't have sufficient data, lead to errors or don't run in time along with a list of their data requirments.

## Usage

Cant(DLM_data, timelimit = 1)

## Arguments

DLM_data A data-limited methods data object (class DLM_data)
timelimit The maximum time (seconds) taken for a method to undertake 10 reps (this filters out methods that are too slow)

CC1 $\quad$| Constant catch management procedure of Geromont and Butterworth |
| :--- |
| (2014) |

## Description

The TAC is the average catch over last yrsmth years.

## Usage

CC1 ( x, DLM_data, $\mathrm{reps}=100, \mathrm{yrsmth}=5, \mathrm{xx}=0$ )

## Arguments

X
DLM_data A data-limited methods data object
reps
The number of TAC samples
yrsmth Years over which to calculate mean catches
$x x \quad$ Parameter controlling the TAC. Mean catches are multiplied by (1-xx)

## Details

This is one of four constant catch rules of Geromont and Butterworth 2014.

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers

## References

Geromont, H.F., Butterworth, D.S. 2014. Generic management procedures for data-poor fisheries; forecasting with few data. ICES J. Mar. Sci. doi:10.1093/icesjms/fst232

CC4 $\quad$| Constant catch management procedure of Geromont and Butterworth |
| :--- |
| $(2014)$ |

## Description

The TAC is the average catch over last yrsmth years reduced by 30

## Usage

CC4 (x, DLM_data, reps $=100, \mathrm{yrsmth}=5, \mathrm{xx}=0.3$ )

## Arguments

| x | A position in data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of TAC samples |
| yrsmth | Years over which to average catches |
| $x x$ | Parameter controlling the TAC. Mean catches are multiplied by (1-xx) |

## Details

This is one of four constant catch MPs of Geromont and Butterworth 2014.

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers

## References

Geromont, H.F., Butterworth, D.S. 2014. Generic management procedures for data-poor fisheries; forecasting with few data. ICES J. Mar. Sci. doi:10.1093/icesjms/fst232

CheckConverg MSE convergence diagnostic

## Description

Have I undertaken enough simulations (nsim)? Has my MSE converged on stable (reliable) peformance metrics?

## Usage

CheckConverg(MSEobj,thresh=2, Plot=TRUE)

## Arguments

MSEobj An object of class 'MSE'
thresh The convergence threshold (percentage). If mean perforamnce metrics are within thresh percent of the second to last interation, the MSE can be considered to have converged.
Plot $\quad$ Should figures be plotted?

## Author(s)

A. Hordyk

ChooseEffort Manually choose the historical relative fishing effort trajectory.

## Description

Interactive plot which allows users to specify the relative trajectory and variability in the historical fishing effort.

## Usage

ChooseEffort(FleetObj, Years=NULL)

## Arguments

| FleetObj | A fleet object. |
| :--- | :--- |
| Years | An optional vector of years. Should be nyears long. |

## Author(s)

A. Hordyk

ChooseSelect Manually choose the historical selectivity pattern.

## Description

Input the first historical year, and all years where selectivity pattern changed (separated by comma).
Interactive plot which allows users to specify a range for the length at $5 \%$ and full selection (LFS), as well as selectivity at maximum length for each year. Produces a simple plot which shows the range in selectivity pattern for each break-point year. Selectivity-at-length is fixed in between breakpoint years. Note that this function replaces 'nyears' in the Fleet object with the value defined here (FstYr:current year).

## Usage

ChooseSelect(Fleet, Stock=NULL, FstYr=NULL, SelYears=NULL)

## Arguments

Fleet A fleet object.
Stock Optional Stock object. If provided, average length-at-maturity is included on plot for reference.
FstYr Optional value for first historical year. If empty, user must specify the year in console.

SelYears Optional vector of values for each year where selectivity pattern changed. If empty, user must specify the years in console (comma separated).

## Author(s)

A. Hordyk
comp Comparison plots for individual simulations

## Description

A simulation by simulation approach to plotting results

## Usage

comp(MSEobj, MPs=NA)

## Arguments

MSEobj An object of class MSE
MPs A character vector of two methods that were applied in making the MSE object

## Author(s)

T. Carruthers

```
CompSRA Age-composition-based estimate of current stock depletion given con-
stant Z linked to an FMSY estimate to provide OFL.
```


## Description

Estimates an OFL based on a Stock Reduction analysis fitted to current age-composition data. Knife-edge vulnerability at age at maturity allows for an FMSY estimate. OFL=FMSY*F/C

## Usage

CompSRA(x, DLM_data, reps = 100)

## Arguments

| $x$ | A position in a data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of stochastic samples of the TAC. |

## Note

Given a fixed historical F, What level of depletion gives you this length composition?

## Author(s)

T. Carruthers

CompSRA4010
Age-composition-based estimate of current stock depletion given constant Z linked to an FMSY estimate to provide OFL (with a 40-10 rule)

## Description

Estimates an OFL based on a Stock Reduction analysis fitted to current age-composition data. Knife-edge vulnerability at age at maturity allows for an FMSY estimate. OFL=FMSY*F/C

## Usage

CompSRA4010(x, DLM_data, reps = 100)

## Arguments

| $x$ | A position in a data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of stochastic samples of the TAC. |

Note
Given a fixed historical F, What level of depletion gives you this length composition?

## Author(s)

T. Carruthers

CSRA $\quad$ Catch at size reduction analysis

## Description

What depletion level and corresponding equlibrium F arise from data regarding mean length of current catches, natural mortality rate, steepness of the stock recruitment curve, maximum length, maximum growth rate, age at maturity, age based vulnerability, maturity at age, maximum age and number of historical years of fishing.

## Usage

CSRA(M, h, Linf, K, t0, AM, a, b, vuln, mat, ML, CAL, CAA, maxage, nyears)

## Arguments

| M | A vector of natural mortality rate estimates |
| :--- | :--- |
| h | A vector of sampled steepness (Beverton-Holt stock recruitment) |
| Linf | A vector of maximum length (von Bertalanffy growth) |
| K | A vector of maximum growth rate (von Bertalanffy growth) |
| t0 | A vector of theoretical age at length zero (von Bertalanffy growth) |
| AM | A vector of age at maturity |
| a | Length-weight conversion parameter a (W=aL^b) |
| b | Length-weight conversion parameter $\mathrm{b}(\mathrm{W}=\mathrm{aL} \wedge \mathrm{b})$ |
| vuln | A matrix nsim x nage of the vulnerabilty at age (max 1) to fishing. |
| mat | A matrix nsim x nage of the maturity at age (max 1) |
| ML | A vector of current mean length estimates |
| CAL | A catch-at-length matrix nyears $x(1$ Linf unit) length bins |
| CAA | A catch-at-age matrix nyears $x$ maximum age |
| maxage | Maximum age |
| nyears | Number of historical years of fishing |

## Author(s)

T. Carruthers
CSRAfunc Optimization function for CSRA

## Description

What depletion level and corresponding equlibrium F arise from data regarding mean length of current catches, natural mortality rate, steepness of the stock recruitment curve, maximum length, maximum growth rate, age at maturity, age based vulnerability, maturity at age, maximum age and number of historical years of fishing.

## Usage

$$
\begin{aligned}
& \text { CSRAfunc (lnF, Mc, hc, maxage, nyears, AFSc, AFCc, Linfc, } \\
& \text { Kc, t0c, AMc, ac, bc, vulnc, matc, MLc, CAL, CAA, } \\
& \text { opt=T, meth="ML") }
\end{aligned}
$$

## Arguments

lnF A proposed value of current instantaneous fishing mortality rate
Mc Natural mortality rate estimates
hc Steepness (Beverton-Holt stock recruitment)
maxage Maximum age
nyears Number of historical years of fishing
AFSc Age at full selection
AFCc Age at first capture
Linfc Maximum length (von Bertalanffy growth)
Kc Maximum growth rate (von Bertalanffy growth)
t0c Theoretical age at length zero (von Bertalanffy growth)
AMc Age at maturity
ac Length-weight conversion parameter a (W=aL^b)
bc Length-weight conversion parameter $b\left(W=a L^{\wedge} b\right)$
vulnc A vector (nage long) of the vulnerabilty at age (max 1) to fishing.
matc A vector (nage long) of the maturity at age (max 1)
MLc A current mean length estimates
CAL A catch-at-length matrix nyears $x$ (1 Linf unit) length bins
CAA A catch-at-age matrix nyears $x$ maximum age
opt Should the measure of fit be returned?
meth Are we fitting to mean length or catch composition?

## Author(s)

T. Carruthers
curE
curE Fishing at current effort levels

## Description

Constant fishing effort set at final year of historical simulations subject to changes in catchability determined by OM@qinc and interannual variability in catchability determined by OM@qcv. This MP is intended to represent a 'status quo' management approach.

## Usage

curE (x, DLM_data, ...)

## Arguments

x
A position in a data-limited methods data object.
DLM_data
$\ldots \quad$ Optional additional arguments that are ignored. Note arguments reps or ... are required for all input controls

## Note

Made up for this package.

## Author(s)

T. Carruthers.
curE75 Fishing at 75 per cent of current effort levels

## Description

Constant fishing effort set at 75 per cent of final year of historical simulations subject to changes in catchability determined by OM@qinc and interannual variability in catchability determined by OM@qcv. This MP is intended to represent a 'status quo' management approach.

## Usage

curE75(x, DLM_data, ...)

## Arguments

x
DLM_data

A position in a data-limited methods data object.
... Optional additional arguments that are ignored. Note arguments reps or ... are required for all input controls

## Note

Made up for this package.

## Author(s)

T. Carruthers.

```
DAAC Depletion Adjusted Average Catch
```


## Description

Essentially DCAC multiplied by $2 *$ depletion and divided by BMSY/B0 (Bpeak)

## Usage <br> DAAC $(x$, DLM_data, reps $=100)$

## Arguments

x A position in a data-limited methods data object

DLM_data
reps The number of stochastic samples of the TAC recommendation

## Author(s)

W. Harford and T. Carruthers

## References

MacCall, A.D., 2009. Depletion-corrected average catch: a simple formula for estimating sustainable yields in data-poor situations. ICES J. Mar. Sci. 66, 2267-2271. Harford W. and Carruthers, T. 2016. Simulation testing novel catch-based fisheries management. In draft, intended for Fish. Bull.

## Description

User prescribed BMSY/B0, M, FMSY/M are used to find B0 and therefore the OFL by backconstructing the stock to match a user specified level of stock depletion (OFL $=\mathrm{M} * \mathrm{FMSY} / \mathrm{M} *$ depletion* B0).

## Usage

DBSRA(x, DLM_data, reps = 100)

## Arguments

$x \quad$ A position in a data-limited methods object.
DLM_data A data-limited methods object.
reps $\quad$ The number of samples of the TAC (OFL) recommendation.

## Details

You specify a range of stock depletion and, given historical catches DB-SRA calculates what unfished biomass must have been to get you here given samples for M, FMSY relative to M and also BMSY relative to Bunfished.

Value
A vector of TAC (OFL) values.

## Note

This is set up to return the OFL (FMSY * current biomass).
You may have noticed that you -the user- specify three of the factors that make the quota recommendation. So this can be quite a subjective method.
Also the DB-SRA method of this package isn't exactly the same as the original method of Dick and MacCall (2011) because it has to work for simulated depletions above BMSY/B0 and even on occasion over B0. Also it doesn't have the modification for flatfish life histories that has previously been applied by Dick and MacCall.

## Author(s)

T. Carruthers

## References

Dick, E.J., MacCall, A.D., 2011. Depletion-Based Stock Reduction Analysis: A catch-based method for determining sustainable yields for data-poor fish stocks. Fish. Res. 110, 331-341.

Depletion-Based Stock Reduction Analysis paired with 40-10 harvest control rule

## Description

User prescribed BMSY/B0, M, FMSY/M are used to find B0 and therefore the OFL by backconstructing the stock to match a user specified level of stock depletion (OFL $=\mathrm{M} *$ FMSY/M * depletion* B0). In this method DBSRA is paried with the 40-10 rule that throttles back the OFL to zero at 10 percent of unfished biomass.

## Usage

DBSRA4010(x, DLM_data, reps = 100)

## Arguments

x
DLM_data A data-limited methods data object
reps The number of stochastic samples of the TAC recommendation

## Author(s)

T. Carruthers

## References

Dick, E.J., MacCall, A.D., 2011. Depletion-Based Stock Reduction Analysis: A catch-based method for determining sustainable yields for data-poor fish stocks. Fish. Res. 110, 331-341.

```
DBSRA_40
Depletion-Based Stock Reduction Analysis assuming 40 per cent stock depletion
```


## Description

DBSRA assuming that current stock depletion is exactly 40 per cent of unfished stock levels.

## Usage

DBSRA_40(x, DLM_data, reps = 100)

## Arguments

X
DLM_data
reps

A position in a data-limited methods data object
A data-limited methods data object
The number of stochastic samples of the TAC recommendation

## Note

A 40 percent assumption for current depletion is more or less the most optimistic state for a stock (ie very close to BMSY/B0 for many stocks).

## Author(s)

T. Carruthers.

## References

Dick, E.J., MacCall, A.D., 2010. Estimates of sustainable yield for 50 data-poor stocks in the Pacific Coast groundfish fishery management plan. Technical memorandum. Southwest fisheries Science Centre, Santa Cruz, CA. National Marine Fisheries Service, National Oceanic and Atmospheric Administration of the U.S. Department of Commerce. NOAA-TM-NMFS-SWFSC-460.

```
DBSRA_ML Depletion-Based Stock Reduction Analysis using mean length estima-
tor of stock depletion
```


## Description

DBSRA using the mean length estimator to calculate current stock depletion.

## Usage

DBSRA_ML(x, DLM_data, reps = 100)

## Arguments

x
DLM_data
reps The number of stochastic samples of the quota recommendation

## Note

The mean length extension was programmed by Gary Nelson as part of his excellent R package 'fishmethods'

## Author(s)

T. Carruthers

## References

Dick, E.J., MacCall, A.D., 2011. Depletion-Based Stock Reduction Analysis: A catch-based method for determining sustainable yields for data-poor fish stocks. Fish. Res. 110, 331-341.
DCAC Depletion Corrected Average Catch

## Description

A method of calculating an MSY proxy (FMSY * BMSY and therefore the OFL at most productive stock size) based on average catches accounting for the windfall catch that got the stock down to BMSY levels.

## Usage

DCAC (x, DLM_data, reps = 100)

## Arguments

| $x$ | A position in a data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of stochastic samples of the TAC recommendation |

Note
It's probably worth noting that DCAC TAC recomemndations do not tend to zero as depletion tends to zero. It adjusts for depletion only in calculating historical average catch. It follows that at stock levels much below BMSY, DCAC tends to chronically overfish.

## Author(s)

T. Carruthers

## References

MacCall, A.D., 2009. Depletion-corrected average catch: a simple formula for estimating sustainable yields in data-poor situations. ICES J. Mar. Sci. 66, 2267-2271.

## DCAC4010 Depletion Corrected Average Catch paired with the 40-10 rule

## Description

A method of calculating an MSY proxy (FMSY * BMSY and therefore the OFL at most productive stock size) based on average catches accounting for the windfall catch that got the stock down to BMSY levels. In this method DCAC is paired with the $40-10$ rule that throttles back the OFL to zero at 10 percent of unfished stock size (the OFL is not subject to downward adjustment above 40 percent unfished)

## Usage

DCAC4010(x, DLM_data, reps = 100)

## Arguments

$x \quad$ A position in a data-limited methods data object
DLM_data A data-limited methods data object
reps The number of stochastic samples of the TAC recommendation

## Note

DCAC can overfish below BMSY levels. The 40-10 harvest control rule largely resolves this problem providing an MP with surprisingly good performance even at low stock levels.

## Author(s)

T. Carruthers

## References

MacCall, A.D., 2009. Depletion-corrected average catch: a simple formula for estimating sustainable yields in data-poor situations. ICES J. Mar. Sci. 66, 2267-2271.

## DCAC_40 <br> Depletion Corrected Average Catch assuming 40 per cent stock deple-

 tion
## Description

DCAC assuming that current stock biomass is exactly 40 per cent of unfished levels.

## Usage

DCAC_40(x, DLM_data, reps = 100)

## Arguments

$x \quad$ A position in a data-limited methods data object
DLM_data A data-limited methods data object
reps $\quad$ The number of stochastic samples of the TAC recommendation

## Note

The 40 percent depletion assumption doesn't really affect DCAC that much as it already makes TAC recommendations that are quite MSY-like.

## Author(s)

T. Carruthers

## References

MacCall, A.D., 2009. Depletion-corrected average catch: a simple formula for estimating sustainable yields in data-poor situations. ICES J. Mar. Sci. 66, 2267-2271.

## DCAC_ML <br> Depletion-Based Stock Reduction Analysis using mean-length estimator of current depletion

## Description

DCAC that uses the mean length estimator to calculate current stock depletion.

## Usage

DCAC_ML(x, DLM_data, reps = 100)

## Arguments

| $x$ | A position in a data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of stochastic samples of the TAC recommendation |

## Note

The mean length extension was programmed by Gary Nelson as part of his excellent R package 'fishmethods'

## Author(s)

T. Carruthers

## References

MacCall, A.D., 2009. Depletion-corrected average catch: a simple formula for estimating sustainable yields in data-poor situations. ICES J. Mar. Sci. 66, 2267-2271.

## Description

A simple delay-difference assessment that estimates the TAC using a time-series of catches and a relative abundance index.

## Usage

DD(x, DLM_data, reps = 100)

## Arguments

$x \quad$ A position in a data-limited methods data object
DLM_data A data-limited methods data object
reps $\quad$ The number of stochastic samples of the TAC recommendation

## Value

A numeric vector of TAC recommendations

## Note

This DD model is observation error only and has does not estimate process error (recruitment deviations). Similar to many other assessment models it depends on a whole host of dubious assumptions such as temporally stationary productivity and proportionality between the abundance index and real abundance. Unsurprisingly the extent to which these assumptions are violated tends to be the biggest driver of performance for this method.

## Author(s)

T. Carruthers

## References

Method based on equations of Carl Walters (bug him with questions and expect colourful responses)

## Description

A simple delay-difference assessment that estimates the OFL using a time-series of catches and a relative abundance index. In this version of the DD MP a 40-10 rule is imposed over the OFL recommendation.

## Usage

DD4010(x, DLM_data, reps = 100)

## Arguments

x
A position in a data-limited methods data object
DLM_data
A data-limited methods data object
reps The number of stochastic samples of the TAC recommendation

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers

## References

Method based on equations of Carl Walters

| DDe | Effort control version of DD - Delay - Difference Stock Assessment <br> with UMSY and MSY leading |
| :--- | :--- |

## Description

A simple delay-difference assessment that estimates and recommends FMSY using a time-series of catches and a relative abundance index.

## Usage

DDe(x, DLM_data, reps = 100)

## Arguments

| $x$ | A position in a data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of stochastic samples of the TAC recommendation |

Note
This DD model is observation error only and has does not estimate process error (recruitment deviations). Similar to many other assessment models it depends on a whole host of dubious assumptions such as temporally stationary productivity and proportionality between the abundance index and real abundance. Unsurprisingly the extent to which these assumptions are violated tends to be the biggest driver of performance for this method.

## Author(s)

T. Carruthers

## References

Method based on equations of Carl Walters (bug him with questions and expect colourful responses)

```
DDe75
```

Effort control version of DD - Delay - Difference Stock Assessment with UMSY and MSY leading that fishes at 75 per cent of FMSY

## Description

A simple delay-difference assessment that estimates and recommends 75 per cent FMSY using a time-series of catches and a relative abundance index.

## Usage

DDe75(x, DLM_data, reps = 100)

## Arguments

| $x$ | A position in a data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of stochastic samples of the TAC recommendation |

## Note

This DD model is observation error only and has does not estimate process error (recruitment deviations). Similar to many other assessment models it depends on a whole host of dubious assumptions such as temporally stationary productivity and proportionality between the abundance index and real abundance. Unsurprisingly the extent to which these assumptions are violated tends to be the biggest driver of performance for this method.

## Author(s)

T. Carruthers

## References

Method based on equations of Carl Walters (bug him with questions and expect colourful responses)
Effort searching version of DD - Delay - Difference Stock Assessment
with UMSY and MSY leading that fishes at 75 per cent of FMSY

## Description

A simple delay-difference assessment that estimates FMSY using a time-series of catches and a relative abundance index. The MP provides a change in effort in the direction of FMSY up to a maximum change of 10 percent.

## Usage

DDes (x, DLM_data, reps $=100, L B=0.9, \quad \mathrm{UB}=1.1$ )

## Arguments

| $x$ | A position in a data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of stochastic samples of the TAC recommendation |
| LB | The lowest permitted factor of previous fishing effort |
| UB | The highest permitted factor of previous fishing effort |

## Note

This DD model is observation error only and has does not estimate process error (recruitment deviations). Similar to many other assessment models it depends on a whole host of dubious assumptions such as temporally stationary productivity and proportionality between the abundance index and real abundance. Unsurprisingly the extent to which these assumptions are violated tends to be the biggest driver of performance for this method.

## Author(s)

T. Carruthers

## References

Method based on equations of Carl Walters (bug him with questions and expect colourful responses)
DepF Depletion Corrected Fratio

## Description

The Fratio MP with a harvest control rule that reduces $F$ according to the production curve given an estimate of current stock depletion.

## Usage

DepF (x, DLM_data, reps = 100)

## Arguments

$x \quad$ A position in data-limited methods data object DLM
DLM_data A data-limited methods data object
reps The number of TAC samples

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers

## References

Made-up for this package.

## DLMdat

Data and Operating model objects

## Description

A dataset including objects for operating models and real data examples

## Usage

DLMdat

## Format

A list of 61 objects

## Description

A way of locating where the package was installed so you can find example data files and code etc.

## Usage

DLMDataDir (stock=NA)

## Arguments

stock Character string representing the name of a .csv file e.g. 'Snapper', 'Rockfish'

## Author(s)

T. Carruthers

```
DLM_data-class Class "DLM_data"
```


## Description

An object for storing data for analysis using data-limited methods

## Objects from the Class

Objects can be created by calls of the form new("DLM_data", stock).

## Slots

Name: The name of the case-study
Year: A vector of years that correspond to catch and relative abundance data
Cat: Total annual catches
Ind: Relative abundance index
$t$ : The number of years corresponding to AvC and Dt
AvC: Average catch over time t
Dt: Depletion over time te.g. Bnow/Bthen
ML: Mean length time series
Mort: Natural mortality rate
FMSY_M: An assumed ratio of FMSY to M
BMSY_B0: The most productive stock size relative to unfished

L50: Length at 50 percent maturity
L95: Length at 95 percent maturity
Lbar: Mean length of catches over Lc (modal length)
Lc: Modal length
LFC: Length at first capture
LFS: smallest Length at full selection
CAA: Catch at Age data
Dep: Stock depletion Bnow/Bunfished (total stock)
Abun: An estimate of absolute current vulnerable abundance
vbK: The von Bertalanffy growth coefficient
vbLinf: Maximum length
vbt0: Theoretical age at length zero
wla: Weight-Length parameter alpha
wlb: Weight-Length parameter beta
steep: Steepness of the Beverton Holt stock-recruitment relationship
CV_Cat: Coefficient of variation in annual catches
CV_Dt: Coefficient of variation in depletion over time $t$
CV_AvC: Coefficient of variation in average catches over time $t$
CV_Ind: Coefficient of variation in the relative abundance index
CV_Mort: Coefficient of variation in natural mortality rate
CV_FMSY_M: Coefficient of variation in the ratio in FMSY/M
CV_BMSY_B0: Coefficient of variation in the position of the most productive stock size relative to unfished
CV_Dep: Coefficient of variation in current stock depletion
CV_Abun: Coefficient of variation in estimate of absolute current stock size
CV_vbK: Coefficient of variation in the von Bert. k parameter
CV_vbLinf: Coefficient of variation in maximum length
CV_vbt0: Coefficient of variation in age at length zero
CV_L50: Coefficient of variation in length at 50 per cent maturity
CV_LFC: Coefficient of variation in length at first capture
CV_LFS: Coefficient of variation in length at full selection
CV_wla: Coefficient of variation in weight-length parameter a
CV_wlb: Coefficient of variation in weight-length parameter $b$
CV_steep: Coefficient of variation in steepness
sigmaL: Assumed observaton error of the length composition data
MaxAge: Maximum age
Units: Units of the catch/absolute abundance estimates

Ref: A reference quota level
Ref_type: Its type
Log: A log of events
params: A place to store estimated parameters
PosMPs: The methods that can be applied to these data
MPs: The methods that were applied to these data
OM: A table of operating model conditions
Obs: A table of observation model conditions
TAC: The calculated TAC
TACbias: The known bias in the calculated TAC
Sense: The results of the sensitivity analysis
CAL_bins: The length bins for the catch-at-length data
CAL: Catch-at-length data
Cref: Reference or target catch level
Iref: Reference or target relative abundance index level
Bref: Reference or target biomass level
CV_Cref: CV for reference or target catch level
CV_Iref: CV for reference or target relative abundance index level
CV_Bref: CV for reference or target biomass level
CV_Rec: CV for recent recruitment strength
Rec: Recent recruitment strength
MPrec: The previous recommendation of a management proceedure
MPeff: The current level of effort
LHYear: The last historical year of the simulation (before projection)
Misc: Optional list which is passed to MPs

## Methods

initialize signature(.Object = "DLM_data"): ...
plot signature(x = "DLM_data"): ...
summary signature(object = "DLM_data"): ...

## Author(s)

T. Carruthers

## Examples

newdata<-new('DLM_data')

```
DLM_fease-class Class "DLM_fease"
```


## Description

An object for storing information about what data are available or might be available

## Objects from the Class

Objects can be created by calls of the form new("DLM_fease", stock).

## Slots

Name: The name of the data feasibility object
Case: The names of the data feasibility cases
Catch: Total annual catches
Index: An index of relative abundance, catch per unit effort data or of fishing mortality rate (effort)
Natural_mortality_rate: From Maximum age, Tagging data, early fishery catch composition data
Maturity_at_length: From gonadal analysis, growth and natural mortality rate estimates
Growth: Paired length and age observations, maximum length and an estimate of natural mortality rate
Length_weight_conversion: Paired weight and length observations, equivalent data from a similar species
Fleet_selectivity: Length composition of catches with growth curve and natural mortality rate, estimates from a similar fleet type targetting a similar species
Catch_at_length: Length composition of catches (length samples)
Catch_at_age: Age composition of catches (age samples)
Recruitment_index: Spawn survey, estimates from a stock assessment, VPA analysis of catch composition data
Stock_recruitment_relationship: Stock assessment, a stock assessment of a similar species
Target_catch: An agreed annual catch target, MSY proxy
Target_biomass: An agreed absolute biomass target, mean historical biomass estimate
Target_index: An agreed catch rate target
Abundance: Fishery independent survey, current fishing mortality rate from recent length composition, natural mortality rate, maturity at age, growth and stock recruitment relationship, habitat and relative density extrapolation

## Methods

initialize signature(.Object = "DLM_fease"): ...
plot signature(x = "DLM_fease"): ...
summary signature(object = "DLM_fease"): ...

## Author(s)

## T. Carruthers

## Examples

```
newdata<-new('DLM_fease')
```

```
DLM_general-class Class "DLM_general"
```


## Description

An object for storing general toolkit data. The data are stored in the right format in the slot data.

## Objects from the Class

Objects can be created by calls of the form new("DLM_general", stock).

## Slots

Name: The name of the data data: The data correctly formated

## Methods

initialize signature(.Object = "DLM_general"): ...
plot signature( $\mathrm{x}=$ "DLM_general"): ...
summary signature(object = "DLM_general"): ...

## Author(s)

T. Carruthers

## Examples

```
newdata<-new('DLM_general')
```

DOM How dominant is an MP?

## Description

The DOM function examines how consistently an MP outperforms another. For example DCAC might provide higher yield than AvC on average but outperforms AvC in less than half of simulations.

## Usage

DOM(MSEobj, MPtg=NA)

## Arguments

MSEobj An object of class 'MSE'
MPtg A character vector of management procedures for cross examination

## Value

A matrix of performance comparisons length(MPtg) rows by MSE@nMPs columns

## Author(s)

A. Hordyk

DoOpt Optimization routine for LBSPR methods

## Description

Internal optimization routine.

## Usage

DoOpt(StockPars, LenDat, SizeBins = NULL, mod = c("GTG", "LBSPR"))

## Arguments

StockPars Life history parameters of stock.
LenDat Binned length data
SizeBins Information on the length bins.
mod Optional for alternative models - only "LBSPR" currently used.

## Author(s)

A. Hordyk

DTe40 Effort searching MP aiming for 40 per cent stock depletion.

## Description

A very simple MP that modifies effort to reach 40 percent stock depletion

## Usage

DTe40(x, DLM_data, reps = 100, alpha=0.4, LB=0.9, UB=1.1)

## Arguments

x
DLM_data
reps
alpha
LB
UB

A position in a data-limited methods data object
A data-limited methods data object
The number of stochastic samples of the TAC recommendation
The target level of depletion
The lowest permitted factor of previous fishing effort
The highest permitted factor of previous fishing effort

## Author(s)

T. Carruthers

## Description

A very simple MP that modifies effort to reach 50 percent stock depletion

## Usage

DTe50(x, DLM_data, reps $=100$, alpha=0.5, LB=0.9, UB=1.1)

## Arguments

X
DLM_data
alpha
LB
UB
reps The number of stochastic samples of the TAC recommendation
A position in a data-limited methods data object
A data-limited methods data object

The target level of depletion
The lowest permitted factor of previous fishing effort
The highest permitted factor of previous fishing effort

## Author(s)

T. Carruthers

DynF
Dynamic Fratio MP

## Description

The Fratio MP with a controller that changes the level of F according to the relationship between Surplus production and biomass. Ie lower $F$ when $d S P / d B$ is positive and higher $F$ when $d S P / d B$ is negative.

## Usage

$\operatorname{DynF}(x$, DLM_data, yrsmth=10, gg=2, reps = 100)

## Arguments

x
DLM_data
yrsmth
gg
reps

A position in a data-limited methods object
A data-limited methods object
The number of historical recent years used for smoothing catch and biomass data
A gain parameter that modifies F according to the gradient in surplus production with biomass

Details
The method smoothes historical catches and biomass and then infers the relationship between surplus production and biomass (as suggested by Mark Maunder and Carl Walters). The approach then regulates a F based policy according to this gradient in which F may range between two different fractions of natural mortality rate.
The core advantage is the $\operatorname{TAC}(\mathrm{t})$ is not strongly determined by $\mathrm{TAC}(\mathrm{t}-1)$ and therefore errors are not as readily propagated. The result is method that tends to perform alarmingly well and therefore requires debunking ASAP.

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers

## References

Made-up for this package.

## Description

Fishing rate is modified each year according to the gradient of surplus production with biomass (aims for zero). F is bounded by FMSY/2 and 2FMSY and walks in the logit space according to $\mathrm{dSP} / \mathrm{dB}$. This is derived from the theory of Maunder 2014.

## Usage

Fadapt(x, DLM_data, reps = 100, yrsmth = 7, gg=1)

## Arguments

x
DLM_data A data-limited methods data object
reps The number of TAC samples
yrsmth Years over which to smooth recent estimates of surplus production
gg
A gain parameter controlling the speed in update in TAC.

## Details

Tested in Carruthers et al. 2015.

## Value

A numeric vector of quota recommendations

## Author(s)

T. Carruthers

## References

Carruthers et al. 2015. Performance evaluation of simple management procedures. Fish and Fisheries. In press. Maunder. 2014. http://www.iattc.org/Meetings/Meetings2014/MAYSAC/PDFs/SAC-05-10b-Management-Strategy-Evaluation.pdf
Fdem Demographic FMSY method

## Description

FMSY is calculated as $r / 2$ where $r$ is calculated from a demographic approach (inc steepness). Coupled with an estimate of current abundance that gives you the OFL.

## Usage

Fdem(x, DLM_data, reps = 100)

## Arguments

$x \quad$ A position in data-limited methods data object
DLM_data A data-limited methods data object
reps The number of TAC samples

## Details

Made up for this package. This uses Murdoch McAllister's demographic r method to derive FMSY (r/2) and then makes the quota r*current biomass / 2. Easy.

## Author(s)

T. Carruthers

## References

McAllister, M.K., Pikitch, E.K., and Babcock, E.A. 2001. Using demographic methods to construct Bayesian priors for the intrinsic rate of increase in the Schaefer model and implications for stock rebuilding. Can. J. Fish. Aquat. Sci. 58: 1871-1890.

| Fdem_CC | Demographic FMSY method using catch-curve analysis to estimate <br> recent $Z$ |
| :--- | :--- |

## Description

FMSY is calculated as $r / 2$ from a demographic $r$ prior method, current abudnance is estimated from naive catch curve analysis.

## Usage

Fdem_CC(x, DLM_data, reps $=100$, Fmin=0.005)

## Arguments

$x \quad$ A position in data-limited methods data object
DLM_data A data-limited methods data object
reps The number of TAC samples
Fmin The minimum fishing mortality rate derived from the catch-curve analysis

## Author(s)

T. Carruthers

## References

McAllister, M.K., Pikitch, E.K., and Babcock, E.A. 2001. Using demographic methods to construct Bayesian priors for the intrinsic rate of increase in the Schaefer model and implications for stock rebuilding. Can. J. Fish. Aquat. Sci. 58: 1871-1890.

## Fdem_ML Demographic FMSY method that uses mean length data to estimate

 recent $Z$
## Description

Demographic F (r/2) method using the mean length estimator to calculate current abundance.

## Usage

Fdem_ML(x, DLM_data, reps = 100)

## Arguments

| $x$ | A position in data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of TAC samples |

## Note

The mean length extension was programmed by Gary Nelson as part of his excellent R package 'fishmethods'

## Author(s)

T. Carruthers

## References

McAllister, M.K., Pikitch, E.K., and Babcock, E.A. 2001. Using demographic methods to construct Bayesian priors for the intrinsic rate of increase in the Schaefer model and implications for stock rebuilding. Can. J. Fish. Aquat. Sci. 58: 1871-1890.
Fease MP feasibility diagnostic

## Description

What MPs may be run (best case scenario) for various data-availability scenarios?

## Usage

Fease(feaseobj,outy="table")

## Arguments

feaseobj An object of class 'DLM_fease'
outy Determines whether you would like a full table or some column of the table for a specific case of the feasibility object. When set equal to table, the full table is produced. When set equal to an integer number the names of MPs that are feasible for that case are returned.

## Author(s)

T. Carruthers

Fease_xl Read in feasibility parameters from Excel spreadsheet

## Description

A function to read in feasibility parameters from an Excel spreadsheet with tabs named following specific convention

## Usage

Fease_xl(fname, stkname, fpath = "", saveCSV = FALSE)

## Arguments

fname Name of the Excel spreadsheet file. Must include file extension.
stkname Name of the Stock.
fpath Full file path, if file is not in current working directory
saveCSV Do you also want to save the Stock, Fleet and Observation parameters to CSV files?

## Details

The Excel spreadsheet must have tabs named with the following convention. For example if stkname is "myFish", the tab must be named "myFishFease,

## Value

A object of class Fease

## Author(s)

## A. Hordyk

## Examples

```
    ## Not run:
    myFease <- Fease_xl(fname="FeaseTables.xlsx", stkname="myFish")
    ## End(Not run)
```

    Fleet-class Class "Fleet"
    
## Description

The component of the operating model that controls fishing dynamics

## Objects from the Class

Objects can be created by calls of the form new("Fleet", OM).

## Slots

Name: Name of the Fleet object
nyears: The number of years for the historical simulation
Spat_targ: Distribution of fishing in relation to spatial biomass: F is proportional to $\mathrm{B}^{\wedge}$ Spat_targ (uniform distribution)
Fsd: Inter-annual variability in fishing mortality rate
EffYears: Vector of verticies, years at which to simulate varying relative effort
EffLower: Lower bound on relative effort corresponding to EffYears (uniform distribution)
EffUpper: Uppper bound on relative effort corresponding to EffYears (uniform distribution)
LFS: Shortest length that is fully vulnerable to fishing (uniform distribution)
L5: Shortest length corresponding of 5 percent vulnerability (uniform distribution)
Vmaxlen: The vulnerability of the longest (oldest) fish (uniform distribution)
SelYears: Vector of verticies, index for years at which historical selectivity pattern changed. If left empty, historical selectivity is constant

AbsSelYears: Optional values for SelYears, used for plotting only. Must be of same length as SelYears
L5Lower: Optional vector of values of length SelYears, specifiying lower limits of L5 (use ChooseSelect function to set these)

L5Upper: Optional vector of values of length SelYears, specifiying upper limits of L5 (use ChooseSelect function to set these)
LFSLower: Optional vector of values of length SelYears, specifiying lower limits of LFS (use ChooseSelect function to set these)
LFSUpper: Optional vector of values of length SelYears, specifiying upper limits of LFS (use ChooseSelect function to set these)

VmaxLower: Optional vector of values of length SelYears, specifiying lower limits of Vmaxlen (use ChooseSelect function to set these)
VmaxUpper: Optional vector of values of length SelYears, specifiying upper limits of Vmaxlen (use ChooseSelect function to set these)
qinc: Average percentage change in fishing efficiency (uniform distribution)(applicable only to forward projection and input controls)
qcv: Inter-annual variability in fishing efficiency (uniform distribution)(applicable only to forward projection and input controls)
isRel: Are the selectivity parameters relative to size-of-maturity? TRUE or FALSE

## Methods

initialize signature(.Object = "Fleet"): ...

## Author(s)

T. Carruthers

## Examples

```
showClass("Fleet")
```


## FMSYref

A reference FMSY method (uses perfect information about FMSY)

## Description

FMSY is taken from the operating model stored at DLM@OM\$FMSY

## Usage

FMSYref(x, DLM_data, reps = 100)

## Arguments

| x | A position in data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of TAC samples |

## Details

Note that you can out-performm this MP even though it has perfect information of FMSY and current abundance. The requirment for fixed $F$ is actually quite strict and is by no means the upper limit in terms of yield. Don't panic if your method beats this one for yield, especially for short-lived species of high temporal variability in productivity!

## Author(s)

T. Carruthers

$$
\begin{aligned}
& \text { FMSYref50 } \\
& \begin{array}{l}
\text { A reference FMSY method that fishes at half of FMSY (uses perfect } \\
\text { information about FMSY) }
\end{array}
\end{aligned}
$$

## Description

FMSY is taken from the operating model stored at DLM@OM\$FMSY

## Usage

FMSYref50(x, DLM_data, reps = 100)

## Arguments

| $x$ | A position in data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of TAC (OFL) samples |

## Details

Note that you can out-performm this method easily. The requirement for fixed F is actually quite strict and is by no means the upper limit in terms of yield. Don't panic if your method beats this one for yield!
Interesting that the reduction in yield is no way near commensurate with the reduction in F - as predicted by a yield curve and expressed in the pretty good yield theory.

## Author(s)

T. Carruthers

FMSYref75 A reference FMSY method that fishes at three quarters of FMSY (uses perfect information about FMSY)

## Description

FMSY is taken from the operating model stored at DLM@OM\$FMSY

## Usage

FMSYref75(x, DLM_data, reps = 100)

## Arguments

x
A position in data-limited methods data object
DLM_data A data-limited methods data object
reps The number of TAC samples

## Details

Note that you can out-performm this method easily. The requirment for fixed F is actually quite strict and is by no means the upper limit in terms of yield. Don't panic if your method beats this one for yield!
Interesting that the reduction in yield is no way near commensurate with the reduction in F as predicted by a yield curve and expressed in the pretty good yield theory.

## Author(s)

T. Carruthers
Fratio An FMSY/M ratio method

## Description

Calculates the OFL based on a fixed ratio of FMSY to M multiplied by a current estimate of abundance.

## Usage

Fratio(x, DLM_data, reps = 100)

## Arguments

| $x$ | A position in a data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of samples of the TAC recommendation |

## Details

A simple method that tends to outperform many other approaches alarmingly often even when current biomass is relatively poorly known. The low stock crash potential is largely due to the quite large difference between Fmax and FMSY for most stocks.

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers

## References

Gulland, J.A., 1971. The fish resources of the ocean. Fishing News Books, West Byfleet, UK.
Martell, S., Froese, R., 2012. A simple method for estimating MSY from catch and resilience. Fish Fish. doi: 10.1111/j.1467-2979.2012.00485.x.

Fratio4010 An FMSY/M ratio method paired with the 40-10 rule

## Description

Calculates the OFL based on a fixed ratio of FMSY to M multiplied by a current estimate of abundance. In this method DBSRA is paired with the 40-10 rule that throttles back the OFL to zero at 10 percent of unfished biomass.

## Usage

Fratio4010(x, DLM_data, reps = 100)

## Arguments

x
DLM_data
reps

A position in data-limited methods data object
A data-limited methods data object
The number of TAC samples

## Author(s)

T. Carruthers

## References

Gulland, J.A., 1971. The fish resources of the ocean. Fishing News Books, West Byfleet, UK.
Martell, S., Froese, R., 2012. A simple method for estimating MSY from catch and resilience. Fish Fish. doi: 10.1111/j.1467-2979.2012.00485.x.

```
Fratio_CC A data-limited method that uses FMSY/M ratio and a naive catch-
    curve estimate of recent Z
```


## Description

Calculates the OFL based on a fixed ratio of FMSY to $M$ and a catch curve estimate of current stock size.

## Usage

Fratio_CC(x, DLM_data, reps = 100, Fmin = 0.005)

## Arguments

| $x$ | A position in data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of TAC samples |
| Fmin | Minimum current fishing mortality rate for the catch-curve analysis |

## Author(s)

T. Carruthers

## References

Gulland, J.A., 1971. The fish resources of the ocean. Fishing News Books, West Byfleet, UK.
Martell, S., Froese, R., 2012. A simple method for estimating MSY from catch and resilience. Fish Fish. doi: 10.1111/j.1467-2979.2012.00485.x.

## Description

Calculates the OFL based on a fixed ratio of FMSY/M and an estimate of current stock size from a mean-length estimator.

## Usage <br> Fratio_ML(x, DLM_data, reps = 100)

## Arguments

x
A position in data-limited methods data object
DLM_data A data-limited methods data object
reps The number of TAC samples

## Note

The mean length extension was programmed by Gary Nelson as part of his excellent R package 'fishmethods'

## Author(s)

T. Carruthers

## References

Gulland, J.A., 1971. The fish resources of the ocean. Fishing News Books, West Byfleet, UK.
Martell, S., Froese, R., 2012. A simple method for estimating MSY from catch and resilience. Fish Fish. doi: 10.1111/j.1467-2979.2012.00485.x.

```
GB_CC
```

Geromont and Butterworth Constant Catch Harvest Control Rule

## Description

A simple MP that aims for average historical catches (as a proxy for MSY) subject to imperfect information.

## Usage

GB_CC(x, DLM_data, reps = 100)

## Arguments

| $x$ | A position in data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of TAC samples |

## Details

Note that this is my interpretation of their MP and is now stochastic. Currently it is generalized and is not 'tuned' to more detailed assessment data which might explain why in some cases it leads to stock declines.

## Author(s)

T. Carruthers

## References

Geromont, H.F. and Butterworth, D.S. 2014. Complex assessment or simple management procedures for efficient fisheries management: a comparative study. ICES J. Mar. Sci. doi:10.1093/icesjms/fsu017

## GB_slope <br> Geromont and Butterworth index slope Harvest Control Rule

## Description

An MP similar to SBT1 that modifies a time-series of catch recommendations and aims for a stable catch rates.

## Usage

GB_slope(x, DLM_data, reps = 100, yrsmth = 5, lambda = 1)

## Arguments

$x \quad$ A position in data-limited methods data object
DLM_data A data-limited methods data object
reps The number of TAC samples
yrsmth Number of years for evaluating slope in relative abundance index
lambda A gain parameter

## Details

Note that this is my interpretation of their approach and is now stochastic. Currently it is generalized and is not 'tuned' to more detailed assessment data which might explain why in some cases it leads to stock declines.

## Author(s)

T. Carruthers

## References

Geromont, H.F. and Butterworth, D.S. 2014. Complex assessment or simple management procedures for efficient fisheries management: a comparative study. ICES J. Mar. Sci. doi:10.1093/icesjms/fsu017
GB_target Geromont and Butterworth target CPUE and catch MP

## Description

An MP similar to SBT2 that modifies a time-series of catch recommendations and aims for target catch rate and catch level based on BMSY/B0 and MSY, respectively.

## Usage

GB_target (x, DLM_data, reps $=100, w=0.5)$

## Arguments

| x | A position in data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of quota samples |
| w | A gain parameter |

## Details

Note that this is my interpretation of their MP and is now stochastic. Currently it is generalized and is not 'tuned' to more detailed assessment data which might explain why in some cases it leads to stock declines.

## Author(s)

T. Carruthers

## References

Geromont, H.F. and Butterworth, D.S. 2014. Complex assessment or simple management procedures for efficient fisheries management: a comparative study. ICES J. Mar. Sci. doi:10.1093/icesjms/fsu017
Gcontrol $\quad G$-control $M P$

## Description

A harvest control rule proposed by Carl Walters that uses trajectory in inferred surplus production to make upward/downward adjustments to TAC recommendations

## Usage

Gcontrol(x, DLM_data, reps = 100, yrsmth $=10$, gg = 2, glim = c(0.5, 2))

## Arguments

x
DLM_data
reps
yrsmth
gg
glim

A position in data-limited methods data object
A data-limited methods data object
The number of quota samples
The number of years over which to smooth catch and biomass data
A gain parameter
A constraint limiting the maximum level of change in quota recommendations

## Author(s)

C. Walters and T. Carruthers

## References

Made-up for this package. Carruthers et al. 2015. Performance of Simple Management Procedures.

## Description

As title.

## Usage

getAFC(t0c, Linfc, Kc, LFC, maxage)

## Arguments

t0c A vector of theoretical age at length zero (von Bertalanffy growth)
Linfc A vector of maximum length (von Bertalanffy growth)
Kc
A vector of maximum growth rate (von Bertalanffy growth)
LFC A vector of length at first capture
maxage Maximum age

## Author(s)

T. Carruthers

Optimization function to find a movement model that matches user specified movement characteristics.

## Description

The user specifies the probability of staying in the same area and spatial heterogeneity (both in the unfished state).

## Usage

getmov(x,Prob_staying,Frac_area_1)

## Arguments

$x \quad$ A position in vectors Prob_staying and Frac_area_1
Prob_staying User specified probability that individuals in area 1 remain in that area (unfished conditions)

Frac_area_1 User specified fraction of individuals found in area 1 (unfished conditions)

## Details

This is paired with movfit to find the correct movement model.

## Value

A markov movement matrix

## Author(s)

T. Carruthers

## Examples

```
Prob_staying<-0.8 # probability that individuals remain in area 1 between time-steps
Frac_area_1<-0.35 # the fraction of the stock found in area 1 under equilibrium conditions
markovmat<-getmov(1,Prob_staying, Frac_area_1)
vec<-c(0.5,0.5) # initial guess at equilibrium distribution (2 areas)
for(i in 1:300)vec<-apply(vec*markovmat,2,sum) # numerical approximation to stable distribution
c(markovmat[1,1],vec[1]) # pretty close right?
```

getq

Optimization function that find the catchability ( $q$ where $F=q E$ ) value required to get to user-specified stock depletion (current biomass / unfished biomass)

## Description

The user specifies the level of stock depleiton. This funciton takes the derived effort trajectories and finds the catchabiltiy to get the stock there.

## Usage

getq(x, dep, Find, Perr, Marray, hs, Mat_age, Wt_age, R0, V, nyears, maxage, mov, Spat_targ, SRrel, $a R, b R$ )

## Arguments

x
dep
Find
Perr
Marray
hs
Mat_age
Wt_age
R0
V
nyears
maxage
mov
Spat_targ
SRrel
aR
bR

## Details

Paired with qopt.

## Author(s)

T. Carruthers
HDAAC Hybrid Depletion Adjusted Average Catch

## Description

Essentially DCAC multiplied by 2*depletion and divided by BMSY/B0 (Bpeak) when below BMSY, and DCAC above BMSY

## Usage

HDAAC(x, DLM_data, reps = 100)

## Arguments

x
DLM_data A data-limited methods data object
reps The number of stochastic samples of the TAC recommendation

## Author(s)

W. Harford and T. Carruthers

## References

MacCall, A.D., 2009. Depletion-corrected average catch: a simple formula for estimating sustainable yields in data-poor situations. ICES J. Mar. Sci. 66, 2267-2271. Harford W. and Carruthers, T. 2016. Testing novel catch-based fisheries management procedures.

```
initialize-methods ~~ Methodsfor Function initialize ~~
```


## Description

~~ Methods for function initialize ~~

## Methods

```
signature(.Object = "DLM")
signature(.Object = "Fleet")
signature(.Object = "MSE")
signature(.Object = "Observation")
signature(.Object = "OM")
signature(.Object = "Stock")
signature(.Object = "lmmodel")
signature(.Object = "DLM_fease")
signature(.Object = "DLM_general")
```

Input Function to run a set of input control methods

## Description

Runs a set of input control methods are returns the output in a single table

## Usage

Input(DLM_data, MPs = NA, reps = 100, timelimit = 10, CheckMPs = TRUE)

## Arguments

| DLM_data | A DLM_data object |
| :--- | :--- |
| MPs | A list of input MPs, if NA all available input MPs are run |
| reps | Number of repetitions (for those methods that use them) |
| timelimit | Maximum timelimit to run MP (in seconds) |
| CheckMPs | Logical, the Can function is run if this is TRUE |

## Author(s)

A. Hordyk

Islope1 A management procedure that incrementally adjusts the TAC to maintain a constant CPUE or relative abundance index.

## Description

The least biologically precautionary of two constant index / CPUE methods proposed by Geromont and Butterworth 2014. Tested by Carruthers et al. 2015

## Usage

Islope1(x, DLM_data, reps = 100, yrsmth = 5, lambda=0.4, xx=0.2)

## Arguments

$x \quad$ A position in data-limited methods data object
DLM_data A data-limited methods data object
reps The number of TAC samples
yrsmth Years over which to smooth recent estimates of surplus production
lambda A gain parameter controlling the speed in update in TAC.
$x x \quad$ Parameter controlling the fraction of mean catch to start using in first year

## Details

Tested by Carruthers et al. 2015.

## Value

A numeric vector of quota recommendations

## Author(s)

T. Carruthers

## References

Carruthers et al. 2015. Performance review of simple management procedures. Fish and Fisheries. In press.
Geromont, H.F., Butterworth, D.S. 2014. Generic management procedures for data-poor fisheries; forecasting with few data. ICES J. Mar. Sci. doi:10.1093/icesjms/fst232

Islope4 A management procedure that incrementally adjusts the TAC to maintain a constant CPUE or relative abundance index.

## Description

The most biologically precautionary of two constant index / CPUE methods proposed by Geromont and Butterworth 2014. Tested by Carruthers et al. 2015

## Usage

Islope4(x, DLM_data, reps = 100, yrsmth = 5, lambda=0.2,xx=0.4)

## Arguments

x
reps The number of TAC samples

DLM_data A data-limited methods data object
yrsmth Years over which to smooth recent estimates of surplus production
lambda A gain parameter controlling the speed in update in TAC.
$x x \quad$ Parameter controlling the fraction of mean catch to start using in first year
A position in data-limited methods data object

## Details

Tested by Carruthers et al. 2015.

## Value

A numeric vector of quota recommendations

## Author(s)

T. Carruthers

## References

Carruthers et al. 2015. Performance evaluation of simple management procedures. Fish and Fisheries. In press.

Geromont, H.F., Butterworth, D.S. 2014. Generic management procedures for data-poor fisheries; forecasting with few data. ICES J. Mar. Sci. doi:10.1093/icesjms/fst232

## IT10 Index Target 10

## Description

An index target MP where the TAC is modified according to current index levels (mean index over last 5 years) relative to a target level. Maximum annual changes are 10 per cent.

## Usage

IT10(x, DLM_data, reps $=100, y r s m t h=5, m c=0.1)$

## Arguments

x
A position in a data-limited methods data object
DLM_data
reps
yrsmth
mc

A data-limited methods data object
The number of stochastic samples of the quota recommendation The number of historical years over which to average the index The maximum fractional change in the TAC among years.

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers

```
IT5
Index Target 5
```


## Description

An index target MP where the TAC is modified according to current index levels (mean index over last 5 years) relative to a target level. Maximum annual changes are 5 per cent.

## Usage

IT5 (x, DLM_data, reps $=100, \mathrm{yrsmth}=5, \mathrm{mc}=0.05$ )

## Arguments

x
DLM_data A data-limited methods data object
reps The number of stochastic samples of the quota recommendation
yrsmth The number of historical years over which to average the index
$m \mathrm{~T} \quad$ The maximum fractional change in the TAC among years.

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers

Itarget1 | A management procedure that incrementally adjusts the TAC (starting |
| :--- |
| from reference level that is a fraction of mean recent catches) to reach |
| a target $C P U E /$ relative abundance index. |

## Description

The least biologically precautionary of two index/CPUE target MPs proposed by Geromont and Butterworth 2014. Tested by Carruthers et al. 2015

## Usage

Itarget1(x, DLM_data, reps = 100, yrsmth = 5, $x x=0$, Imulti=1.5)

## Arguments

x
DLM_data
reps
yrsmth Years over which to smooth recent estimates of surplus production
$x x \quad$ Parameter controlling the fraction of mean catch to start using in first year
Imulti Parameter controlling how much larger target CPUE / index is compared with recent levels.

## Details

Tested by Carruthers et al. 2015.

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers

## References

Carruthers et al. 2015. Performance evaluation of simple management procedures. Fish and Fisheries. In press.
Geromont, H.F., Butterworth, D.S. 2014. Generic management procedures for data-poor fisheries; forecasting with few data. ICES J. Mar. Sci. doi:10.1093/icesjms/fst232

Itarget4 A management procedure that incrementally adjusts the TAC (starting from reference level that is a fraction of mean recent catches) to reach a target CPUE / relative abundance index.

## Description

The most biologically precautionary of two index/CPUE target MPs proposed by Geromont and Butterworth 2014. Tested by Carruthers et al. 2015

## Usage

Itarget4(x, DLM_data, reps = 100, yrsmth = 5, xx=0.3, Imulti=2.5)

## Arguments

x
DLM_data
reps
yrsmth
xx
Imulti

A position in data-limited methods data object
A data-limited methods data object
The number of TAC samples
Years over which to smooth recent estimates of surplus production
Parameter controlling the fraction of mean catch to start using in first year
Parameter controlling how much larger target CPUE / index is compared with recent levels.

## Details

Tested by Carruthers et al. 2015.

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers

## References

Carruthers et al. 2015. Performance evaluation of simple management procedures. Fish and Fisheries. In press.

Geromont, H.F., Butterworth, D.S. 2014. Generic management procedures for data-poor fisheries; forecasting with few data. ICES J. Mar. Sci. doi:10.1093/icesjms/fst232

ItargetE1 $\quad$| A management procedure that incrementally adjusts the effort to reach |
| :--- |
| a target $C P U E /$ relative abundance index. |

## Description

An effort-based version of the least biologically precautionary of two index/CPUE target MPs proposed by Geromont and Butterworth 2014. Tested by Carruthers et al. 2015

## Usage

ItargetE1(x, DLM_data, reps $=100, \mathrm{yrsmth}=5, \mathrm{xx}=0$, $\operatorname{Imulti}=1.5$ )

## Arguments

X
DLM_data
reps
yrsmth
xx
Imulti Parameter controlling how much larger target CPUE / index is compared with recent levels.

## Details

Tested by Carruthers et al. 2015.

## Value

A numeric vector of input controls

## Author(s)

T. Carruthers

## References

Carruthers et al. 2015. Performance evaluation of simple management procedures. Fish and Fisheries. In press.
Geromont, H.F., Butterworth, D.S. 2014. Generic management procedures for data-poor fisheries; forecasting with few data. ICES J. Mar. Sci. doi:10.1093/icesjms/fst232

ItargetE4 | A management procedure that incrementally adjusts the Effort to reach |
| :--- |
| a target $C P U E /$ relative abundance index. |

## Description

An effort-based version of the most biologically precautionary of two index/CPUE target MPs proposed by Geromont and Butterworth 2014.

## Usage

ItargetE4(x, DLM_data, reps = 100, yrsmth = 5, xx = 0, Imulti = 2.5)

## Arguments

$x \quad$ A position in data-limited methods data object
DLM_data A data-limited methods data object
reps The number of samples
yrsmth Years over which to smooth recent estimates of surplus production
$x x \quad$ Parameter controlling the fraction of mean catch to start using in first year
Imulti Parameter controlling how much larger target CPUE / index is compared with recent levels.

## Details

Tested by Carruthers et al. 2015.

## Value

A numeric vector of input controls

## Author(s)

T. Carruthers

## References

Carruthers et al. 2015. Performance evaluation of simple management procedures. Fish and Fisheries. In press.
Geromont, H.F., Butterworth, D.S. 2014. Generic management procedures for data-poor fisheries; forecasting with few data. ICES J. Mar. Sci. doi:10.1093/icesjms/fst232

## ITe10 Index Target Effort-Based 10

## Description

An index target MP where the Effort is modified according to current index levels (mean index over last 5 years) relative to a target level. Maximum annual changes are 10 per cent.

## Usage

ITe10(x, DLM_data, reps $=100$, yrsmth $=5, \mathrm{mc}=0.1$ )

## Arguments

x
DLM_data
reps
yrsmth
mc
reps

A position in a data-limited methods data object

A data-limited methods data object
The number of stochastic samples of the quota recommendation The number of historical years over which to average the index The maximum fractional change in the Effort among years.

## Value

A numeric vector of input controls

## Author(s)

T. Carruthers

$$
\text { ITe5 } \quad \text { Index Target Effort-Based } 5
$$

## Description

An index target MP where the Effort is modified according to current index levels (mean index over last 5 years) relative to a target level. Maximum annual changes are 5 per cent.

## Usage

ITe5(x, DLM_data, reps $=100$, yrsmth $=5, \mathrm{mc}=0.05)$

## Arguments

X
DLM_data
reps
yrsmth
mc

A position in a data-limited methods data object
A data-limited methods data object
The number of stochastic samples of the quota recommendation
The number of historical years over which to average the index
The maximum fractional change in the effort among years.

## Value

A numeric vector of input controls

## Author(s)

T. Carruthers

## Description

An index target MP where the TAC is modified according to current index levels (mean index over last yrsmth years) relative to a target level. Maximum fractional annual changes are mc. $\mathrm{mc}=(5+\mathrm{M} * 25) / 100 \mathrm{yrsmth}=4^{*}(1 / \mathrm{M})^{\wedge}(0.25)$

## Usage

ITM(x, DLM_data, reps = 100)

## Arguments

| $x$ | A position in a data-limited methods data object |
| :--- | :--- |
| DLM_data | A data-limited methods data object |
| reps | The number of stochastic samples of the quota recommendation |

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers
joinMSE Join multiple MSE objects together

## Description

Joins two or more MSE objects together. MSE objects must have identical number of historical years, and projection years.

## Usage

joinMSE(MSEobjs = NULL)

## Arguments

MSEobjs A list of MSE objects. Must all have identical operating model and MPs. MPs which don't appear in all MSE objects will be dropped.

## Value

An object of class MSE

## Author(s)

A. Hordyk

## KalmanFilter

Kalman filter and Rauch-Tung-Striebel smoother

## Description

Kalman filter to predict new points and smoother for time-series.

## Usage

KalmanFilter(RawEsts, $\mathrm{R}=1, \mathrm{Q}=0.1$, Int $=100$ )

## Arguments

RawEsts Vector of numeric values to be filtered and smoothed.
$R \quad$ Variance of sampling noise
Q Variance of random walk increments
Int Covariance of initial uncertainty

## Author(s)

A. Hordyk

Kplot
KOBE plot: a projection by projection plot of F/FMSY and B/BMSY

## Description

A standard KOBE plot by each method that also shows the percentage of methods that ended up in each quadrant.

## Usage

Kplot(MSEobj,maxsim=60, nam=NA)

## Arguments

| MSEobj | An object of class MSE |
| :--- | :--- |
| maxsim | Maximum number of simulations (lines) to plot on each panel. |
| nam | The name of the plot |

## Note

Apologies for the nauseating shading.

## Author(s)

T. Carruthers

L2A
Length to age conversion

## Description

Simple deterministic length to age conversion given inverse von Bertalanffy growth.

## Usage

L2A(t0c, Linfc, Kc, Len, maxage)

## Arguments

| t0c | Theoretical age at length zero |
| :--- | :--- |
| Linfc | Maximum length |
| Kc | Maximum growth rate |
| Len | Length |
| maxage | Maximum age |

## Value

An age (vector of ages, matrix of ages) corresponding with Len

## Author(s)

T. Carruthers

## LBSPR

Apply LBSPR model to time-series of catch-at-length

## Description

Apply LBSPR model to time-series of catch-at-length data, and return smoothed estimates of SPR, and relative fishing mortality (F/M).

## Usage

LBSPR(x, DLM_data, yrsmth = 1, reps = reps)

## Arguments

x
DLM_data DLM_data object
yrsmth Number of years to smooth length data - currently not used
reps Currently not used

## Author(s)

## A. Hordyk

## References

Hordyk, A.R., Ono, K., Sainsbury, K.J., Loneragan, N., and Prince, J.D. 2015. Some explorations of the life history ratios to describe length composition, spawning-per-recruit, and the spawning potential ratio. ICES J. Mar. Sci. 72: 204-216.

Hordyk, A.R., Ono, K., Valencia, S.R., Loneragan, N.R., and Prince, J.D. 2015. A novel lengthbased empirical estimation method of spawning potential ratio (SPR), and tests of its performance, for small-scale, data-poor fisheries. ICES J. Mar. Sci. 72: 217-231.

## Description

Function that takes stock (biology) and fleet (selectivity and fishing mortality) parameters and returns expected equilibrum size strucure of the catch and spawning potential ratio.

## Usage

LBSPRSim(StockPars, FleetPars, SizeBins = NULL, P = 0.001, Nage = 101)

## Arguments

StockPars The life-history parameters of the stock.
FleetPars $\quad$ The relative fishing mortality $(\mathrm{F} / \mathrm{M})$ and the selectivity-at-length parameters.
SizeBins Optional - specifies the maximum length bin (ToSize) and the length increment of the binned length data (Linc). If empty, Linc $=5$ and ToSize $=\operatorname{Linf}+$ MaxSD * SDLinf
$P \quad$ Percentage of initial cohort still alive at maximum age.
Nage Maximum age of generic age-strucured model. Not sensitive, unless set too low (e.g., < 50).

## Author(s)

A. Hordyk

## References

Hordyk, A.R., Ono, K., Sainsbury, K.J., Loneragan, N., and Prince, J.D. 2015. Some explorations of the life history ratios to describe length composition, spawning-per-recruit, and the spawning potential ratio. ICES J. Mar. Sci. 72: 204-216.
Hordyk, A.R., Ono, K., Valencia, S.R., Loneragan, N.R., and Prince, J.D. 2015. A novel lengthbased empirical estimation method of spawning potential ratio (SPR), and tests of its performance, for small-scale, data-poor fisheries. ICES J. Mar. Sci. 72: 217-231.

## Description

Iteratively adjusts Effort based on distance between estimated and target SPR (40\%), and slope of recent SPR estimates.

## Usage

LBSPR_ItEff(x, DLM_data, yrsmth = 1, reps = reps)

## Arguments

x
Simulation number
DLM_data
DLM_data object
yrsmth Number of years to smooth length data - not currently used
reps Not currently used

## Author(s)

A. Hordyk

LBSPR_ItSel Length-based SPR model with HCR that iteratively adjusts Selectivity.

## Description

Management Procedure which adjusts size-at-selection based on estimated SPR. Entirely untested, and included at to demonstrate MPs of this type.

## Usage

LBSPR_ItSel(x, DLM_data, yrsmth = 1, reps = reps)

## Arguments

x
DLM_data
yrsmth Number of years to smooth length data - not currently used
reps
Not currently used

## Author(s)

A. Hordyk

## Description

Iteratively adjusts TAC based on distance between estimated and target SPR (40\%), and slope of recent SPR estimates.

## Usage

LBSPR_ItTAC (x, DLM_data, yrsmth = 1, reps = reps)

## Arguments

x
DLM_data
yrsmth Number of years to smooth length data - not currently used
reps Not currently used

## Author(s)

A. Hordyk
lmmodel-class Class "lmmodel"

## Description

An object for storing fitted linear model objects in this case the relationship between M , age-atmaturity and the von B. K parameter.

## Objects from the Class

Objects can be created by calls of the form new("lmmodel", stock).

## Slots

Name: The Name of the list of linear models
models: A list of fitted linear models

## Methods

initialize signature(.Object = "lmmodel"): ...

## Author(s)

T. Carruthers

## Examples

newdata<-new('lmmodel',"Name", new('list'))

```
LstepCC1
```

A management procedure that incrementally adjusts the TAC according to the mean length of recent catches.

## Description

The least biologically precautionary of four adaptive length-based MPs proposed by Geromont and Butterworth 2014. Tested by Carruthers et al. 2015

## Usage

LstepCC1 (x, DLM_data, reps $=100, \mathrm{yrsmth}=5, \mathrm{xx}=0$, stepsz=0.05, llim=c(0.96,0.98,1.05))

## Arguments

X
DLM_data
reps
yrsmth
XX
stepsz
llim

A position in data-limited methods data object
A data-limited methods data object
The number of TAC samples
Years over which to smooth recent estimates of surplus production
Parameter controlling the fraction of mean catch to start using in first year
Parameter controlling the size of the TAC update increment.
A vector of length reference points that determine the conditions for increasing, maintaining or reducing the TAC.

## Details

Tested by Carruthers et al. 2015.

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers

## References

Carruthers et al. 2015. Performance evaluation of simple management procedures. Fish and Fisheries. In press.

Geromont, H.F., Butterworth, D.S. 2014. Generic management procedures for data-poor fisheries; forecasting with few data. ICES J. Mar. Sci. doi:10.1093/icesjms/fst232

LstepCC4 A management procedure that incrementally adjusts the TAC according to the mean length of recent catches.

## Description

The most biologically precautionary of four adaptive length-based MPs proposed by Geromont and Butterworth 2014. Tested by Carruthers et al. 2015

## Usage

LstepCC4 (x, DLM_data, reps $=100, y r s m t h=5, x x=0.3$, stepsz=0.05, llim=c $(0.96,0.98,1.05)$ )

## Arguments

x
DLM_data A data-limited methods data object
reps The number of TAC samples
yrsmth Years over which to smooth recent estimates of surplus production
$x x \quad$ Parameter controlling the fraction of mean catch to start using in first year
stepsz Parameter controlling the size of the TAC update increment.
llim A vector of length reference points that determine the conditions for increasing, maintaining or reducing the TAC.

## Details

Tested by Carruthers et al. 2015.

## Value

A numeric vector of TAC recommendations

## Author(s)

T. Carruthers

## References

Carruthers et al. 2015. Performance evaluation of simple management procedures. Fish and Fisheries. In press.

Geromont, H.F., Butterworth, D.S. 2014. Generic management procedures for data-poor fisheries; forecasting with few data. ICES J. Mar. Sci. doi:10.1093/icesjms/fst232

```
LstepCE1 A management procedure that incrementally adjusts the TAC accord-
``` ing to the mean length of recent catches.

\section*{Description}

A effort-based version of least biologically precautionary of four adaptive length-based MPs proposed by Geromont and Butterworth 2014. Tested by Carruthers et al. 2015

\section*{Usage}

LstepCE1 (x, DLM_data, reps \(=100\), yrsmth \(=5, x x=0\), stepsz \(=0.05\), llim \(=c(0.96,0.98,1.05))\)

\section*{Arguments}

X
DLM_data A data-limited methods data object
reps The number of effort samples
yrsmth
\(x x \quad\) Parameter controlling the fraction of mean catch to start using in first year
stepsz Parameter controlling the size of the effort update increment.
llim A vector of length reference points that determine the conditions for increasing, maintaining or reducing the effort.

\section*{Value}

A numeric vector of input controls

\section*{Author(s)}
T. Carruthers

LstepCE2 A management procedure that incrementally adjusts the Effort according to the mean length of recent catches.

\section*{Description}

A effort-based version of one of the four adaptive length-based MPs proposed by Geromont and Butterworth 2014.

\section*{Usage}

LstepCE2(x, DLM_data, reps = 100, yrsmth = 5, xx = 0, stepsz = 0.1, llim \(=c(0.96,0.98,1.05))\)

\section*{Arguments}
x
A position in data-limited methods data object
DLM_data A data-limited methods data object
reps The number of samples
yrsmth Years over which to smooth recent estimates of surplus production
\(x x \quad\) Parameter controlling the fraction of mean catch to start using in first year
stepsz Parameter controlling the size of the effort update increment.
llim A vector of length reference points that determine the conditions for increasing, maintaining or reducing the effort.

\section*{Value}

A numeric vector of input controls

\section*{Author(s)}
T. Carruthers
\begin{tabular}{ll} 
Ltarget1 & \begin{tabular}{l} 
A management procedure that incrementally adjusts the TAC to reach \\
a target mean length in catches.
\end{tabular}
\end{tabular}

\section*{Description}

The least biologically precautionary of four target length MPs proposed by Geromont and Butterworth 2014. Tested by Carruthers et al. 2015

\section*{Usage}

Ltarget1(x, DLM_data, reps = 100, yrsmth = 5, xx=0, xL=1.05)

\section*{Arguments}
x
A position in data-limited methods data object
DLM_data A data-limited methods data object
reps The number of TAC samples
yrsmth Years over which to smooth recent estimates of surplus production
\(x x \quad\) Parameter controlling the fraction of mean catch to start using in first year
\(x \mathrm{~L} \quad\) Parameter controlling the magnitude of the target mean length of catches relative to average length in catches.

\section*{Details}

Tested by Carruthers et al. 2015.

\section*{Value}

A numeric vector of TAC recommendations

\section*{Author(s)}
T. Carruthers

\section*{References}

Carruthers et al. 2015. Performance evaluation of simple management procedures. Fish and Fisheries. In press.
Geromont, H.F., Butterworth, D.S. 2014. Generic management procedures for data-poor fisheries; forecasting with few data. ICES J. Mar. Sci. doi:10.1093/icesjms/fst232

Ltarget4 A management procedure that incrementally adjusts the TAC to reach a target mean length in catches.

\section*{Description}

The most biologically precautionary of four target length MPs proposed by Geromont and Butterworth 2014. Tested by Carruthers et al. 2015

\section*{Usage}

Ltarget4(x, DLM_data, reps \(=100, \mathrm{yrsmth}=5, \mathrm{xx}=0.2, \mathrm{xL}=1.15\) )

\section*{Arguments}
x
DLM_data A data-limited methods data object
reps The number of TAC samples
yrsmth Years over which to smooth recent estimates of surplus production
\(x x \quad\) Parameter controlling the fraction of mean catch to start using in first year
\(x \mathrm{~L} \quad\) Parameter controlling the magnitude of the target mean length of catches relative to average length in catches.

\section*{Details}

Tested by Carruthers et al. 2015.

\section*{Value}

A numeric vector of TAC recommendations

\section*{Author(s)}
T. Carruthers

\section*{References}

Carruthers et al. 2015. Performance evaluation of simple management procedures. Fish and Fisheries. In press.

Geromont, H.F., Butterworth, D.S. 2014. Generic management procedures for data-poor fisheries; forecasting with few data. ICES J. Mar. Sci. doi:10.1093/icesjms/fst232
LtargetE1
\begin{tabular}{l} 
A management procedure that incrementally adjusts the Effort to reach \\
a target mean length in catches.
\end{tabular}

\section*{Description}

A effort based version of the least biologically precautionary of four target length MPs proposed by Geromont and Butterworth 2014.

\section*{Usage}

LtargetE1 (x, DLM_data, reps \(=100, \mathrm{yrsmth}=5, \mathrm{xx}=0, \mathrm{xL}=1.05\) )

\section*{Arguments}
x
DLM_data
reps
yrsmth
xx
xL

A position in data-limited methods data object
A data-limited methods data object
The number of samples
Years over which to smooth recent estimates of surplus production
Parameter controlling the fraction of mean catch to start using in first year
Parameter controlling the magnitude of the target mean length of catches relative to average length in catches.

\section*{Value}

A numeric vector of input controls

\section*{Author(s)}
T. Carruthers
\begin{tabular}{ll} 
LtargetE4 & \begin{tabular}{l} 
A management procedure that incrementally adjusts the Effort to reach \\
a target mean length in catches.
\end{tabular}
\end{tabular}

\section*{Description}

A effort based version of the most biologically precautionary of four target length MPs proposed by Geromont and Butterworth 2014.

\section*{Usage}

LtargetE4(x, DLM_data, reps = 100, yrsmth \(=5, \mathrm{xx}=0, \mathrm{xL}=1.15\) )

\section*{Arguments}
\begin{tabular}{ll}
x & A position in data-limited methods data object \\
DLM_data & A data-limited methods data object \\
reps & The number of samples \\
yrsmth & Years over which to smooth recent estimates of surplus production \\
xx & Parameter controlling the fraction of mean catch to start using in first year \\
xL & \begin{tabular}{l} 
Parameter controlling the magnitude of the target mean length of catches relative \\
to average length in catches.
\end{tabular}
\end{tabular}

\section*{Value}

A numeric vector of input controls

\section*{Author(s)}
T. Carruthers

\section*{Description}

Takes an existing OM object and converts it to one without any observation error, and very little process error. Used for debugging and testing that MPs perform as expected under perfect conditions.

\section*{Usage}
makePerf(0Min, except \(=\) NULL)

\section*{Arguments}

OMin An object of class OM
except An optional vector of slot names in the OM that will not be changed (not tested perfectly so watch out!)

\section*{Value}

A new OM object

\section*{Author(s)}
A. Hordyk
matlenlim \begin{tabular}{l} 
A data-limited method in which fishing vulnerability is set according \\
to the maturity curve
\end{tabular}

\section*{Description}

An example of the implementation of input controls in the DLM toolkit, where selectivity-at-length is set equivalent to maturity-at-length

\section*{Usage}
matlenlim(x, DLM_data, ...)

\section*{Arguments}
\begin{tabular}{ll}
\(x\) & A position in a data-limited methods object \\
DLM_data & A data-limited methods object \\
\(\ldots\) & Optional additional arguments that are ignored. Note arguments reps or \(\ldots\) \\
& are required for all input controls
\end{tabular}

\section*{Value}

A vector of input control recommendations, with values for length at first capture and full selection

\section*{Author(s)}
T. Carruthers

\section*{References}

Made-up for this package
matlenlim2 \(\quad\)\begin{tabular}{l} 
A data-limited method in which fishing vulnerability is set slightly \\
higher than the maturity curve
\end{tabular}

\section*{Description}

An example of the implementation of input controls in the DLM toolkit, where selectivity-at-length is set slightly higher than the maturity-at-length

\section*{Usage}
matlenlim2(x, DLM_data, ...)

\section*{Arguments}
x
DLM_data
... Optional additional arguments that are ignored. Note arguments reps or ... are required for all input controls

\section*{Value}

A vector of input control recommendations, with values for length at first capture and full selection

\section*{Author(s)}
A. Hordyk

\section*{References}

Made-up for this package

\section*{Description}

A simple average catch-depletion MP that was included to demonstrate just how informative an estimate of current stock depletion can be. TAC \(=2 * D * A v C\)

\section*{Usage}

MCD (x, DLM_data, reps = 100)

\section*{Arguments}
\(x \quad\) A position in a data-limited methods data object
DLM_data A data-limited methods data object
reps The number of stochastic samples of the quota recommendation

\section*{Value}

A numeric vector of TAC recommendations

\section*{Author(s)}
T. Carruthers
MCD4010 Mean Catch Depletion

\section*{Description}

A simple average catch-depletion MP linked to a 40-10 harvest controle rule that was included to demonstrate just how informative an estimate of current stock depletion can be. TAC=d(1-d)AvC

\section*{Usage}

MCD4010(x, DLM_data, reps = 100)

\section*{Arguments}
\(x \quad\) A position in a data-limited methods data object
DLM_data A data-limited methods data object
reps The number of stochastic samples of the quota recommendation

Value
A numeric vector of TAC recommendations

\section*{Author(s)}

\section*{T. Carruthers}

Depletion and F estimation from mean length of catches

\section*{Description}

A highly dubious means of getting very uncertain estimates of current stock biomass and (equilibrium) fishing mortality rate from growth, natural mortality rate, recruitment and fishing selectivity.

\section*{Usage}
ML2D(OM,ML, nsim=100,ploty=T,Dlim=c(0.05,0.6))

\section*{Arguments}

OM
An object of class 'OM'
ML A estimate of current mean length of catches
nsim Number of simulations
ploty \(\quad\) Produce a plot of depletion and F
Dlim Limits on the depletion that is returned as a fraction of unfished biomass.

\section*{Value}

A table of nsim rows and 2 columns (depletion, fishing mortality rate)

\section*{Author(s)}
T. Carruthers

\section*{Description}

The user specifies the probability of staying in the same area and spatial heterogeneity (both in the unfished state). This function returns the squared difference between these values and those produced by the three logit movement model.

\section*{Usage}
movfit(par, prb,frac)

\section*{Arguments}
par Three parameters in the logit space that control the four probabilities of moving between 2 areas
prb User specified probability that individuals in area 1 remain in that area (unfished conditions)
frac User specified fraction of individuals found in area 1 (unfished conditions)

\section*{Details}

This is paired with getmov to find the correct movement model.

\section*{Author(s)}
T. Carruthers
```

MRnoreal
An marine reserve in area 1 with no spatial reallocation of fishing
effort

```

\section*{Description}

A spatial control that prevents fishing in area 1 and does not reallocate this fishing effort to area 2.

\section*{Usage}

MRnoreal(x, DLM_data, ...)

\section*{Arguments}
x
DLM_data

A position in data / simulation object DLM
A data limited methods data object
Optional additional arguments that are ignored. Note arguments reps or ... are required for all input controls

\section*{Author(s)}
T. Carruthers
\[
\text { MRreal } \quad \text { An marine reserve in area } 1 \text { with full reallocation of fishing effort }
\]

\section*{Description}

A spatial control that prevents fishing in area 1 and reallocates this fishing effort to area 2.

\section*{Usage}

MRreal(x, DLM_data, ...)

\section*{Arguments}
x
DLM_data

A position in data / simulation object DLM
A data limited methods data object
Optional additional arguments that are ignored. Note arguments reps or ... are required for all input controls

\section*{Author(s)}
T. Carruthers
MSE-class Class "MSE"

\section*{Description}

A Management Strategy Evaluation object that contains information about simulation conditions and performance of data-limited methods

\section*{Objects from the Class}

Objects can be created by calls of the form new("MSE", Name, nyears, proyears, nMPs, MPs, nsim, OMtable, Obs, B

\section*{Slots}

Name: Name of the MSE object
nyears: The number of years for the historical simulation
proyears: The number of years for the projections - closed loop simulations
nMPs: Number of management procedures simulation tested
MPs: The names of the MPs that were tested
nsim: Number of simulations
OM: A table of nsim rows with a column for each sampled parameter of the operating model
- RefY: reference yield, the highest long-term yield (mean over last five years of projection) obtained from a fixed F strategy. This is a useful reference point for framing performance of MPs because it standardizes for starting point and future productivity.
- M: instantaneous natural mortality rate
- Depletion: stock depletion (biomass / unfished biomass) in the final historical year (prior to projection)
- A: abundance (biomass) updated in each management update of projection
- BMSY_B0: most productive stock size relative to unfished
- FMSY_M: fishing mortality rate divided by natural mortality rate
- Mgrad: mean average percentage gradient in natural mortality rate (percentage per time step)
- Msd: interannual variability in natural mortality rate (lognormal CV)
- procsd: process error - CV in log-normal recruitment deviations
- Esd: interannual variability in historical effort (fishing mortality rate)
- dFfinal: gradient in fishing mortality rate over final five years of the historical simulation
- MSY: Maximum Sustainable Yield
- qinc: mean percentage increase in fishing efficiency (catchability) in projected years (input controls only)
- qcv: interannual variability in future fishing efficiency (catchability) in projected years (input controls only)
- CALcv: variability in lengths at age around the growth curve (normal CV)
- FMSY: Fishing mortality rate at Maximum Sustainable Yield
- Linf: maximum length (von Bertalanffy Linf parameter)
- K: maximum growth rate (von Bertalanffy K parameter)
- t0: theoretical length at age zero (von Bertalanffy t0 parameter)
- hs: steepness of the stock recruitment relationship (the fraction of unfished recruitment at a fifth of unfished stock levels)
- Linfgrad: mean gradient in maximum length (per cent per time step)
- Kgrad: mean gradient in maximum growth rate (per cent per time step)
- Linfsd: interannual variability in maximum length (log normal CV)
- recgrad: gradient in recruitment strength (age 1 population numbers) over last 10 years of historical simulations
- Ksd: interannual variability in maximum growth rate (log normal CV)
- ageM: age at 50 per cent maturity
- LFS: length at full selection (the shortest length class where fishery selectivity is 100 per cent)
- age05: the age at 5 percent selectivity (ascending limb of selectivity curve)
- Vmaxage: the selectivity of the oldest age class (controls dome shape of selectivity curve)
- LFC: length at first capture, the smallest length that can be caught by the gear
- OFLreal: the true simulated Over Fishing Limit (FMSY x biomass) updated in each management update of the projection
- Spat_targ: spatial targetting parameter, fishing mortality rate across areas is proportional to vulnerable biomass raised to the power of this number.
- Frac_area_1: the fraction of unfished biomass inhabiting area 1 (can be seen as fraction of habitat in area 1 or relative size of area 1)
- Prob_staying: the probability that individuals in area 1 remain there between time-steps
- AC: autocorrelation in recruitment

Obs: A table of nsim rows with a column for each sampled parameter of the observation model
- Cbias: bias in observed catches
- Csd: observation error in observed catches (lognormal CV)
- CAA_nsamp: the number of catch-at-age observations per time step
- CAA_ESS: the effective sample size of multinomial catch-at-age observation model (number of independent draws)
- CAL_nsamp: the number of catch-at-length observations per time step
- CAL_ESS: the effective sample size of multinomial catch-at-length observation model (number of independent draws)
- Isd: observation error in relative abundance index (lognormal CV)
- Dbias: bias in observed stock depletion (also applies to depletion Dt for DCAC)
- Mbias: bias in observed natural mortality rate
- FMSY_Mbias: bias in ratio of FMSY to natural mortality rate
- BMSY_B0bias: bias in ratio of most productive stock size relative to unfished
- AMbias: bias in age at 50 per cent maturity
- LFCbias: bias in length at first capture
- LFSbias: bias in length at full selection
- Abias: bias in observed current absolute stock biomass
- Kbias: bias in maximum growth rate (von Bertalanffy K parameter)
- t0bias: bias in theoretical length at age zero (von Bertalanffy t0 parameter)
- Linfbias: bias in maximum length (von Bertalanffy Linf parameter)
- hbias: bias in observed steepness of the stock recruitment relationship
- Irefbias: bias in abundance index corresponding to BMSY stock levels
- Crefbias: bias in MSY prediction (target or reference catch)
- Brefbias: bias in BMSY stock levels (target or reference biomass levels)

B_BMSY: Stored biomass relative to BMSY over the projection (an array with dimensions nsim, nMPs, proyears)

F_FMSY: Stored fishing mortality rate relative to FMSY over the projection (an array with dimensions nsim, nMPs, proyears)
B: Stored stock biomass over the projection (an array with dimensions nsim, nMPs, proyears)
FM: Stored fishing mortality rate over the projection (an array with dimensions nsim, nMPs, proyears)
C: Stored catches (taken) over the projection (an array with dimensions nsim, nMPs, proyears)
TAC: Stored Total Allowable Catch (prescribed) over the projection (an array with dimensions nsim, nMPs, proyears)(note that this is NA for input controls)
SSB_hist: Stored historical spawning stock biomass (historical simulations - an array with dimensions nsim, nages, nyears, nareas)
CB_hist: Stored historical catches in weight (historical simulations - an array with dimensions nsim, nages, nyears, nareas)
FM_hist: Stored historical fishing mortality rate (historical simulations - an array with dimensions nsim, nages, nyears, nareas)

\section*{Methods}
initialize signature(.Object = "MSE"):
plot signature(x = "MSE"):

\section*{Author(s)}
T. Carruthers
NAor0 Is a value NA or zero.

\section*{Description}

As title

\section*{Usage}

NAor0(x)

\section*{Arguments}
\(x \quad\) A numeric value.

\section*{Value}

A table of nsim rows and 2 columns (depletion, fishing mortality rate)

\section*{Author(s)}
T. Carruthers

Needed Data needed to get MPs running

\section*{Description}

Wrapper function for DLMdiag that lists what data are needed to run data-limited methods that are current not able to run given a DLM_cdata object

\section*{Usage}

Needed(DLM_data, timelimit=1)

\section*{Arguments}
\begin{tabular}{ll} 
DLM_data & A data-limited methods data object \\
timelimit & The maximum time (seconds) taken to complete 10 reps
\end{tabular}

\section*{Author(s)}
T. Carruthers
NFref No Fishing Reference MP

\section*{Description}

A reference MP that sets annual catch to zero (or very close to it). Used for looking at variability in stock with no fishing.

\section*{Usage}

NFref(x, DLM_data, reps = 100)

\section*{Arguments}
x
DLM_data
reps The number of stochastic samples of the quota recommendation

\section*{Value}

A TAC of 0.01

\section*{Author(s)}
A. Hordyk

NOAA_plot
National Oceanographic and Atmospheric Administration default plot 1

\section*{Description}

A preliminary plot for returning trade-offs plots and performance table for total yield, variability in yield, probability of overfishing and likelihood of biomass dropping below 50 per cent BMSY

\section*{Usage}

NOAA_plot(MSEobj, nam=NA, type=NA, panel=T)

\section*{Arguments}

MSEobj An object of class MSE
nam Title of plot
type Plots full range of data if NA. Plots a subset that meet thresholds if not NA.
panel Should a two panel plot be made or should plots be made in sequence.

\section*{Value}

A table of performance metrics.

\section*{Author(s)}
T. Carruthers

\section*{Description}

An operating model component that controls the observation model

\section*{Objects from the Class}

Objects can be created by calls of the form new("Observation", OM).

\section*{Slots}

Name: The name of the observation model object
Cobs: Log-normal catch observation error expressed as a coefficient of variation (uniform distribution)

Cbiascv: A coefficient of variation controlling the sampling of bias in catch observations for each simulation (uniform distribution)
CAA_nsamp: Number of catch-at-age observation per time step (uniform distribution)
CAA_ESS: Effective sample size (independent age draws) of the multinomial catch-at-age observation error model (uniform distribution)
CAL_nsamp: Number of catch-at-length observation per time step (uniform distribution)
CAL_ESS: Effective sample size (independent length draws) of the multinomial catch-at-length observation error model (uniform distribution)
CALcv: Lognormal, variability in the length at age (uniform distribution)
Iobs: Observation error in the relative abundance indices expressed as a coefficient of variation (uniform distribution)
Mcv: Persistent bias in the prescription of natural mortality rate sampled from a log-normal distribution with coefficient of variation (Mcv)(uniform distribution)
Kcv: Persistent bias in the prescription of growth parameter k sampled from a log-normal distribution with coefficient of variation (Kcv)(uniform distribution)
t0cv: Persistent bias in the prescription of t 0 sampled from a log-normal distribution with coefficient of variation (t0cv)(uniform distribution)
Linfcv: Persistent bias in the prescription of maximum length sampled from a log-normal distribution with coefficient of variation (Linfcv)(uniform distribution)
LFCcv: Persistent bias in the prescription of lenght at first capture sampled from a log-normal distribution with coefficient of variation (LFCcv)(uniform distribution)
LFScv: Persistent bias in the prescription of length-at-fully selection sampled from a log-normal distribution with coefficient of variation (LFScv)(uniform distribution)
B 0 cv : Persistent bias in the prescription of maximum lengthunfished biomass sampled from a lognormal distribution with coefficient of variation ( B 0 cv )(uniform distribution)
FMSYcv: Persistent bias in the prescription of FMSY sampled from a log-normal distribution with coefficient of variation (FMSYcv)(uniform distribution)
FMSY_Mcv: Persistent bias in the prescription of FMSY/M sampled from a log-normal distribution with coefficient of variation (FMSY_cv)(uniform distribution)
BMSY_B0cv: Persistent bias in the prescription of BMsY relative to unfished sampled from a lognormal distribution with coefficient of variation (BMSY_B0cv)(uniform distribution)
rcv: Persistent bias in the prescription of intrinsic rate of increase sampled from a log-normal distribution with coefficient of variation (rcv)(uniform distribution)
LenMcv: Persistent bias in the prescription of length at 50 percent maturity sampled from a lognormal distribution with coefficient of variation (A50cv)(uniform distribution)
Dbiascv: Persistent bias in the prescription of stock depletion sampled from a log-normal distribution with coefficient of variation (Linfcv)(uniform distribution)

Dcv: Imprecision in the prescription of stock depletion among years, expressed as a coefficient of variation (uniform distribution)

Btbias: Persistent bias in the prescription of current stock biomass sampled from a uniform-log distribution with range (Btbias)(uniform distribution)
Btcv: Imprecision in the prescription of current stock biomass among years expressed as a coefficient of variation (uniform distribution)

Fcurbiascv: Persistent bias in the prescription of current fishing mortality rate sampled from a log-normal distribution with coefficient of variation (Fcurcv)(uniform distribution)
Fcurcv: Imprecision in the prescription of current fishing mortality rate among years expressed as a coefficient of variation (uniform distribution)
hcv : Persistent bias in steepness (uniform distribution)
Icv: Observation error in realtive abundance index expressed as a coefficient of variation (uniform distirbution)
maxagecv: Bias in the prescription of maximum age (uniform distribution)
beta: A parameter controlling hyperstability/hyperdepletion. I^beta therefore values below 1 lead to hyperstability (an index that decreases slower than true abundance) and values above 1 lead to hyperdepletion (an index that decreases more rapidly than true abundance)(uniform distribution)

Reccv: Bias in the knowledge of recent recruitment strength (uniform distribution)
Irefcv: Bias in the knowledge of the relative abundance index at BMSY (uniform distribution)
Brefcv: Bias in the knowledge of BMSY (uniform distribution)
Crefcv: Bias in the knowledge of MSY(uniform distribution)

\section*{Methods}
initialize signature(.Object = "Observation"): ...

\section*{Note}

Its questionable whether the hyperstability/hyperdepletion should be categorised as an observation model characteristic as it is most often driven by fleet dynamics (and therefore should be in the fleet object). Oh well its here and you might want to make it hyperstable beta \(<1\) or hyperdeplete beta \(>\) 1, only.

\section*{Author(s)}
T. Carruthers

\section*{Examples}
showClass("Observation")

OM-class Class "OM"

\section*{Description}

An object containing all the parameters needed to control the MSE which can be build from component Stock, Fleet and Observation objects. Almost all of these inputs are a vector of length 2 which describes the upper and lower bounds of a uniform distribution from which to sample the parameter.

\section*{Objects from the Class}

Objects can be created by calls of the form new("OM", Stock, Fleet, Observation).

\section*{Slots}

Name: Name of the operating model
nyears: The number of years for the historical simulation
maxage: The maximum age of individuals that is simulated (there is no 'plus group': individuals die off beyone the maximum age so there isn't a huge cost to simulating more older age classes)
R0: The magnitude of unfished recruitment. This is normally fixed to some arbitrary value since it simply scales the simulated numbers)
M: Natural mortality rate (uniform distribution)
Msd: Inter-annual variability in natural mortality rate expressed as a coefficient of variation (uniform distribution)
Mgrad: Mean temporal trend in natural mortality rate, expressed as a percentage change in M per year (uniform distribution)
h: Steepness of the stock recruit relationship (uniform distribution)
SRrel: Type of stock-recruit relationship (1)Beverton-Holt (2) Ricker
Linf: Maximum length (uniform distribution)
K: von B. growth parameter k (uniform distribution)
t0: von B. theoretical age at length zero (uniform distribution)
Ksd: Inter-annual variability in growth parameter k (uniform distribution)
Kgrad: Mean temporal trend in growth parameter k , expressed as a percentage change in k per year (uniform distribution)
Linfsd: Inter-annual variability in maximum length - uniform distribution
Linfgrad: Mean temporal trend in maximum length, expressed as a percentage change in Linf per year (uniform distribution)
recgrad: Mean temporal trend in log-normal recruitment deviations (uniform distribution)
AC : Autocorrelation in recruitment deviations \(\operatorname{rec}(\mathrm{t})=\mathrm{AC}{ }^{*} \operatorname{rec}(\mathrm{t}-1)+(1-\mathrm{AC}) * \operatorname{sigma}(\mathrm{t})\) (uniform distribution)
a: Length-weight parameter alpha (uniform distribution)
b: Length-weight parameter beta (uniform distribution)
D: Current level of stock depletion (Bcurrent/Bunfished) (uniform distribution)
Size_area_1: The size of area 1 relative to area 2 (uniform distribution)
Frac_area_1: The fraction of the unfished biomass in stock 1 (uniform distribution)
Prob_staying: The probability of inviduals in area 1 remaining in area 1 over the course of one year
Source: A reference to a website or article form which parameters were taken to define the operating model
beta: A parameter controlling hyperstability/hyperdepletion. I^beta therefore values below 1 lead to hyperstability (an index that decreases slower than true abundance) and values above 1 lead to hyperdepletion (an index that decreases more rapidly than true abundance)(uniform distribution)
Spat_targ: Distribution of fishing in relation to spatial biomass: F is proportional to \(\mathrm{B}^{\wedge}\) Spat_targ (uniform distribution)

LFS: Shortest length that is fully vulnerable to fishing (uniform distribution)
L5: Shortest length at 5 percent vulnerability (uniform distribution)
Vmaxlen: The vulnerability of the longest (oldest) fish (uniform distribution)
SelYears: Vector of verticies that index years where historical selectivity pattern changed. Leave empty to ignore
AbsSelYears: vector of absolute year values that correspond to year indices in SelYears. Used only for plotting
L5Lower: Optional vector of values of length SelYears, specifiying lower limits of L5 (use ChooseSelect function to set these. Overrides L5 above)

L5Upper: Optional vector of values of length SelYears, specifiying upper limits of L5 (use ChooseSelect function to set these. Overrides L5 above)
LFSLower: Optional vector of values of length SelYears, specifiying lower limits of LFS (use ChooseSelect function to set these. Overrides LFS above)

LFSUpper: Optional vector of values of length SelYears, specifiying upper limits of LFS (use ChooseSelect function to set these. Overrides LFS above)
VmaxLower: Optional vector of values of length SelYears, specifiying lower limits of Vmaxlen (use ChooseSelect function to set these. Overrides Vmaxlen above)

VmaxUpper: Optional vector of values of length SelYears, specifiying upper limits of Vmaxlen (use ChooseSelect function to set these. Overrides Vmaxlen above)
isRel: Are the selectivity parameters relative to size-of-maturity? TRUE or FALSE
L50: Length at 50 percent maturity (uniform distribution)
L50_95: Length increment from 50 to 95 percent maturity (uniform distribution)
Fsd: Inter-annual variability in fishing mortality rate
EffYears: Vector of verticies, years at which to simulate varying relative effort
EffLower: Lower bound on relative effort corresponding to EffYears (uniform distribution)

EffUpper: Uppper bound on relative effort corresponding to EffYears (uniform distribution)
qinc: Average percentage change in fishing efficiency (uniform distribution)(applicable only to forward projection and input controls)
qcv: Inter-annual variability in fishing efficiency (uniform distribution)(applicable only to forward projection and input controls)
Cobs: Log-normal catch observation error expressed as a coefficient of variation (uniform distribution)

Cbiascv: A coefficient of variation controlling the sampling of bias in catch observations for each simulation (uniform distribution)
CAA_nsamp: Number of catch-at-age observation per time step (uniform distribution)
CAA_ESS: Effective sample size (independent age draws) of the multinomial catch-at-age observation error model (uniform distribution)
CAL_nsamp: Number of catch-at-length observation per time step (uniform distribution)
CAL_ESS: Effective sample size (independent length draws) of the multinomial catch-at-length observation error model (uniform distribution)
CALcv: Lognormal, variability in the length at age (uniform distribution)
Iobs: Observation error in the relative abundance indices expressed as a coefficient of variation (uniform distribution)
Perr: The extent of inter-annual log-normal recruitment variability (sigma R)(uniform distribution)
Period: Period for cylical recruitment pattern in years (uniform distribution). Leave empty to ignore

Amplitude: Amplitude in deviation from long-term average recruitment during recruitment cycle, both positive and negative (uniform distribution). E.g., a range from 0 to 0.5 means recruitment decreases or increases by up to \(50 \%\) each cycle. Leave empty to ignore
Mcv: Persistent bias in the prescription of natural mortality rate sampled from a log-normal distribution with coefficient of variation (Mcv)(uniform distribution)

Kcv : Persistent bias in the prescription of growth parameter k sampled from a log-normal distribution with coefficient of variation (Kcv)(uniform distribution)
t0cv: Persistent bias in the prescription of t0 sampled from a log-normal distribution with coefficient of variation (t0cv)(uniform distribution)
Linfcv: Persistent bias in the prescription of maximum length sampled from a log-normal distribution with coefficient of variation (Linfcv)(uniform distribution)

LFCcv: Persistent bias in the prescription of lenght at first capture sampled from a log-normal distribution with coefficient of variation (LFCcv)(uniform distribution)
LFScv: Persistent bias in the prescription of length-at-fully selection sampled from a log-normal distribution with coefficient of variation (LFScv)(uniform distribution)

B 0 cv : Persistent bias in the prescription of maximum lengthunfished biomass sampled from a lognormal distribution with coefficient of variation (B0cv)(uniform distribution)
FMSYcv: Persistent bias in the prescription of FMSY sampled from a log-normal distribution with coefficient of variation (FMSYcv)(uniform distribution)
FMSY_Mcv: Persistent bias in the prescription of FMSY/M sampled from a log-normal distribution with coefficient of variation (FMSY_cv)(uniform distribution)

BMSY_B0cv: Persistent bias in the prescription of BMsY relative to unfished sampled from a lognormal distribution with coefficient of variation (BMSY_B0cv)(uniform distribution)
\(r c v:\) Persistent bias in the prescription of intrinsic rate of increase sampled from a log-normal distribution with coefficient of variation (rcv)(uniform distribution)

LenMcv: Persistent bias in the prescription of length at 50 percent maturity sampled from a lognormal distribution with coefficient of variation (A50cv)(uniform distribution)

Dbiascv: Persistent bias in the prescription of stock depletion sampled from a log-normal distribution with coefficient of variation (Linfcv)(uniform distribution)

Dcv: Imprecision in the prescription of stock depletion among years, expressed as a coefficient of variation (uniform distribution)

Btbias: Persistent bias in the prescription of current stock biomass sampled from a uniform-log distribution with range (Btbias)(uniform distribution)

Btcv: Imprecision in the prescription of current stock biomass among years expressed as a coefficient of variation (uniform distribution)
Fcurbiascv: Persistent bias in the prescription of current fishing mortality rate sampled from a log-normal distribution with coefficient of variation (Fcurcv)(uniform distribution)

Fcurcv: Imprecision in the prescription of current fishing mortality rate among years expressed as a coefficient of variation (uniform distribution)
hcv: Persistent bias in steepness (uniform distribution)
Icv: Observation error in realtive abundance index expressed as a coefficient of variation (uniform distirbution)
maxagecv: Bias in the prescription of maximum age (uniform distribution)
Reccv: Bias in the knowledge of recent recruitment strength (uniform distribution)
Irefcv: Bias in the knowledge of the relative abundance index at BMSY (uniform distribution)
Brefcv: Bias in the knowledge of BMSY (uniform distribution)
Crefcv: Bias in the knowledge of MSY(uniform distribution)

\section*{Methods}
initialize signature(. Object = "OM"): ...

\section*{Author(s)}
T. Carruthers

\section*{Examples}
showClass("OM")

OM_xl Read in operating model parameters from Excel spreadsheet

\section*{Description}

A function to read in operating model parameters from an Excel spreadsheet with tabs named following specific convention.

\section*{Usage}

OM_xl(fname, stkname, fpath = "", saveCSV = FALSE)

\section*{Arguments}
\begin{tabular}{ll} 
fname & Name of the Excel spreadsheet file. Must include file extension. \\
stkname & Name of the Stock. \\
fpath & Full file path, if file is not in current working directory \\
saveCSV & \begin{tabular}{l} 
Do you also want to save the Stock, Fleet and Observation parameters to CSV \\
files?
\end{tabular}
\end{tabular}

\section*{Details}

The Excel spreadsheet must have tabs named with the following convention. For example if stkname is "myFish", the Stock parameters are in a tab named "myFishStock"", Fleet parameters in a tab named "myFishFleet", and Observation parameters in a tab named "myFishObs". All three tabs (Stock, Fleet and Obs) must be present for a single stock. You can have multiple stocks in a single spreadsheet, provided that the stock names are different.

\section*{Value}

A object of class OM

\section*{Author(s)}

\section*{A. Hordyk}

\section*{Examples}
```


## Not run:

OM <- OM_xl(fname="OMTables.xlsx", stkname="myFish")

## End(Not run)

```

OptFun Objective function for LBSPR methods

\section*{Description}

Used internally to define objective function for LBSPR methods.

\section*{Usage}

OptFun(tryFleetPars, LenDat, StockPars, SizeBins = NULL, mod = c("GTG", "LBSPR"))

\section*{Arguments}
tryFleetPars Vector of relative fishing mortality (F/M) and selectivity-at-length parameters.
LenDat Binned length data
StockPars Life history parameters of stock.
SizeBins Information on the length bins.
mod Optional for alternative models - only "LBSPR" currently used.

\section*{Author(s)}
A. Hordyk
ourMSE Example MSE object used in the vignette

\section*{Description}

A dummy example MSE object, with customized stock and fleet, all MPs, and 16 simulations.

\section*{Usage}
data("ourMSE")

\section*{Format}

The format is: Formal class 'MSE' [package "DLMtool"] with 17 slots ..@ Name : chr "Stock:Snapper Fleet:Generic_FlatE Observation model:Imprecise_Biased" ..@ nyears : num 50 ..@ proyears: num 20 ..@ nMPs : int 63 ..@ MPs : chr [1:63] "AvC" "BK" "BK_CC" "BK_ML" ... ..@ nsim : num 16 ..@ OM :'data.frame': 16 obs. of 34 variables: .. ..\$ RefY : num [1:16] 50.2 120.150 .867 .8 78.4 ... .. ..\$ M : num [1:16] 0.2130 .2390 .2170 .2430 .241 ... .. ..\$ Depletion : num [1:16] 0.311 0.2570 .2570 .1990 .13 ... .. ..\$ A : num [1:16] 144.4656 .6191 .972 .2146 .1 ... .. ..\$ BMSY_B0 : num [1:16] 0.3250 .3250 .3250 .2960 .306 ... .. ..\$ FMSY_M : num [1:16] 0.5790 .5640 .5560 .664 0.62 ... .. ..\$ Mgrad : num [1:16] -0.1677-0.0985-0.1874 0.1456 0.0794 ... .. .. \$ Msd : num [1:16] \(0.03130 .02530 .0150 .03640 .011 \ldots\).. .. \(\$\) procsd : num [1:16] 0.4750 .4150 .3310 .3430 .436
... .. ..\$ Esd : num [1:16] 0.2930 .130 .3720 .1060 .312 ... .. ..\$ dFfinal : num [1:16] -0.009183 -0.007048-0.006787-0.003407-0.000569 ... .. .. \$ MSY : num [1:16] 55.251 .139 .248 .751 .8 ... .. ..\$ qinc : num [1:16] -0.686-0.127 1.558 0.780 .692 ... .. .. \$ qcv : num [1:16] 0.1590 .2980 .187 0.2410 .102 ... .. ..\$ FMSY : num [1:16] 0.1230 .1350 .1210 .1610 .149 ... .. ..\$ Linf : num [1:16] 91.189 .788 .991 .289 .9 ... .. .. \(\$ \mathrm{~K}\) : num [1:16] 0.1980 .2050 .1790 .1990 .21 ... .. .. \(\$ \mathrm{t} 0\) : num [1:16] -0.0367-0.0296-0.0333-0.0322-0.0319 ... .. ..\$ hs : num [1:16] 0.8560 .8750 .8560 .937 0.94 ... .. ..\$ Linfgrad : num [1:16] -0.185 0.0209-0.1721-0.185-0.1004 ... .. ..\$ Kgrad : num [1:16] -0.0662 0.06670 .14380 .07260 .1547 ... .. .. \(\$\) Linfsd : num [1:16] 0.013340 .012030 .02384 0.013770 .00404 ... .. .. \$ recgrad : num [1:16] -0.731 \(2.751-4.075-0.234-1.097\)... .. ..\$ Ksd : num [1:16] 0.004430 .011850 .014740 .009210 .02128 ... .. .. \$ ageM : num [1:16] 2.261 .972 .25 2.512 .25 ... .. ..\$ V26 : num [1:16] 12.510 .310 .813 .711 .1 ... .. ..\$ V27 : num [1:16] 3227 31.834 .830 .8 ... .. ..\$ V28 : num [1:16] 0.7980 .6810 .8960 .6870 .902 ... .. ..\$ LFC : num [1:16] 13.413 .310 .912 .112 .6 ... .. .. \(\$\) OFLreal : num [1:16] 51.428 .431 .14022 .3 ... .. .. \(\$\) Spat_targ : num [1:16] 1111111111 ... .. ..\$ Frac_area_1: num [1:16] 0.1360 .1380 .1450 .070 .133 ... .. ..\$ Prob_staying: num [1:16] 0.8840 .9110 .6510 .4270 .658 ... .. ..\$ AC : num [1:16] 0.569 0.7170 .7630 .5720 .8 ... ..@ Obs :'data.frame': 16 obs. of 25 variables: .. .. \(\$\) Cbias : num [1:16] 0.6591 .0070 .7451 .8661 .27 ... .. ..\$ Csd : num [1:16] 0.2020 .3110 .5540 .5750 .589 ... .. ..\$ CAA_nsamp : num [1:16] 88598671978187809799 ... .. ..\$ CAA_ESS : num [1:16] 13 191819131219151518 ... .. .. \(\$\) CAL_nsamp : num [1:16] 77.171 .272 .377 .874 ... .. .. \(\$\) CAL_ESS : num [1:16] 19191617151915141419 ... .. ..\$ Isd : num [1:16] 0.5120 .2890 .439 0.5430 .352 ... .. ..\$ Dbias : num [1:16] 0.6212 .0230 .7651 .3250 .648 ... .. ..\$ Derr : num [1:16] 0.06630 .06470 .09210 .17880 .1261 ... .. .. \$ Mbias : num [1:16] 1.1081 .7691 .2710 .9010 .397 ... .. ..\$ FMSY_Mbias: num [1:16] 1.2080 .981 .1892 .058 0.583 ... .. ..\$ BMSY_B0bias: num [1:16] 0.8970 .9470 .9980 .9420 .953 ... .. .. \(\$\) lenMbias : num [1:16] 1.0870 .8981 .0670 .8210 .842 ... .. ..\$ LFCbias : num [1:16] 0.9390 .960 .8961 .0420 .903 ... .. ..\$ LFSbias : num [1:16] 0.977 1.0060 .9261 .2520 .76 ... .. .. \(\$\) Abias : num [1:16] 0.3312 .9870 .2720 .2311 .854 ... .. .. \(\$\) Aerr : num [1:16] 0.4490 .2710 .2620 .3020 .39 ... .. ..\$ Kbias : num [1:16] 0.8991 .1661 .031 .1570 .841 ... .. .. \(\$\) t0bias : num [1:16] 1.01910 .9820 .9541 .096 ... .. ..\$ Linfbias : num [1:16] 0.9250 .927 0.9291 .0380 .971 ... .. .. \(\$\) hbias : num [1:16] 1.0210 .9771 .0290 .9830 .969 ... .. .. \(\$\) Irefbias : num [1:16] 1.2640 .7181 .2810 .9020 .952 ... .. .. \(\$\) Crefbias : num [1:16] 0.8390 .5760 .9210 .9081 .1 ... .. ..\$ Brefbias : num [1:16] 0.6551 .1411 .0950 .7251 .58 ... .. ..\$ betas : num [1:16] 0.3890 .521 0.3661 .6890 .357 ... ..@ B_BMSY : num [1:16, 1:63, 1:20] 1.096 0.4720 .9560 .8280 .371 ... ..@ F_FMSY : num [1:16, 1:63, 1:20] 0.4471 .3440 .8022 .2142 .207 ... ..@ B : num [1:16, 1:63, 1:20] 579223346305144 ... ..@ FM : num [1:16, 1:63, 1:20] 0.05520 .18160 .09690 .35660 .3297 ... ..@ C : num [1:16, 1:63, 1:20] 29.53833 .5116 .354 ... ..@ TAC : num [1:16, 1:63, 1:20] 29.538 33.5116 .354 ... ..@ SSB_hist: num [1:16, 1:34, 1:50, 1:2] 0.01250 .05130 .04630 .01640 .0113 ... ..@ CB_hist : num [1:16, 1:34, 1:50, 1:2] 0000000000 ... ..@ FM_hist : num [1:16, 1:34, 1:50, 1:2] \(0000000000 \ldots\)

\section*{Examples}
data(ourMSE)
ourReefFish Example data object

\section*{Description}

Example data object with a number of output control MPs run on it, and includes resulting distributions of TACs

\section*{Usage \\ data("ourReefFish")}

\section*{Format}

The format is: Formal class 'DLM_data' [package "DLMtool"] with 71 slots ..@ Name : chr "ourReefFish" ..@ Year : num [1:55] 19541955195619571958 ... ..@ Cat : num [1, 1:55] 1.33 1.461 .661 .562 .23 ... ..@ Ind : num [1, 1:55] 0.1630 .1380 .1480 .1570 .146 ... ..@ Rec : num [1, 1:55] NA NA NA NA NA NA NA NA NA NA ... ..@ t : num 54 ..@ AvC : num 1.84 ..@ Dt : num 0.545 ..@ Mort : num 0.094 ..@ FMSY_M : num 0.7 ..@ BMSY_B0 : num 0.3 ..@ Cref : num NA ..@ Bref : num NA ..@ Iref : num NA ..@ L50 : num 31.6 ..@ L95 : num 52.5 ..@ LFC : num 20 ..@ LFS : num 36 ..@ CAA : num [1, 1:21, 1:48] 0000000000 ... ..@ Dep : num 0.2 ..@ Abun : num NA ..@ vbK : num 0.192 ..@ vbLinf : num 85.6 ..@ vbt0 : num -0.395 ..@ wla : num 0.0000167 ..@ wlb : num 2.95 ..@ steep : num 0.99 ..@ CV_Cat : num 0.2 ..@ CV_Dt : num 0.25 ..@ CV_AvC : num 0.2 ..@ CV_Ind : num 0.2 ..@ CV_Mort : num 0.2 ..@ CV_FMSY_M : num 0.2 ..@ CV_BMSY_B0: num 0.045 ..@ CV_Cref : num 0.2 ..@ CV_Bref : num 0.2 ..@ CV_Iref : num 0.2 ..@ CV_Rec : num 0.2 ..@ CV_Dep : num 0.25 ..@ CV_Abun : num 0.25 ..@ CV_vbK : num 0.00803 ..@ CV_vbLinf : num 0.00325 ..@ CV_vbt0 : num 0.00777 ..@ CV_L50 : num 0.1 ..@ CV_LFC : num 0.2 ..@ CV_LFS : num 0.2 ..@ CV_wla : num 0.1 ..@ CV_wlb : num 0.1 ..@ CV_steep : num 0.1 ..@ sigmaL : num 0.2 ..@ MaxAge : num 48 ..@ Units : chr "thousand tonnes" ..@ Ref : num NA ..@ Ref_type : chr NA ..@ Log :List of 1 .. ..\$ : chr "Created: 2016-02-18 14:28:11" ..@ params : list() ..@ PosMPs : chr [1:31] "AvC" "BK_CC" "BK_ML" "CC1" ... ..@ MPs : chr [1:31] "AvC" "BK_CC" "BK_ML" "CC1" ... ..@ OM :'data.frame': 1 obs. of 1 variable: .. ..\$ NA.: logi NA ..@ Obs :'data.frame': 1 obs. of 1 variable: .. ..\$ NA.: logi NA ..@ TAC : num [1:31, 1:100, 1] 1.3615 .911 .143 .081 .77 ... ..@ TACbias : logi [1, 1, 1] NA ..@ Sense : logi \([1,1,1]\) NA ..@ CAL_bins : num [1:44] 0202224262830323436 ... ..@ CAL : num \([1,1: 28,1: 43] 0021010000 \ldots\)..@ MPrec : num NA ..@ MPeff : num NA ..@ ML : num [1, 1:55] NA NA NA NA NA NA NA NA NA NA ... ..@ Lbar : num [1, 1:55] NA NA NA NA NA NA NA NA NA NA ... ..@ Lc : num [1, 1:55] NA NA NA NA NA NA NA NA NA NA ... ..@ LHYear : num 2008 ..@ Misc : list()

\section*{Examples}
```

data(ourReefFish)
str(ourReefFish) ;
plot(ourReefFish)

```
plot-methods \(\quad \sim \sim\) Methods for Function plot \(\sim \sim\)

\section*{Description}
~~Methods for function plot ~~

\section*{Methods}
```

    signature(x = "DLM_data")
    signature(x = "MSE")
    ```
    plot0FL A generic OFL plot for NOAA use

\section*{Description}

As title.

\section*{Usage}
plotOFL(DLM_data, xlims=NA, perc=0.5)

\section*{Arguments}
\begin{tabular}{ll} 
DLM_data & An object of class DLM_data that has been run though TAC() \\
xlims & x axis limits \\
perc & The percentile of the OFL distribution to be plotted
\end{tabular}

\section*{Value}

A table of performance metrics.

\section*{Author(s)}
T. Carruthers
\begin{tabular}{l} 
PorgMSE \(\quad\) Example MSE object used in the vignette \\
\hline
\end{tabular}

\section*{Description}

A dummy example MSE object, based on porgy, generic fleet and imprecise and unbiased observation model, four new MPs, and 20 simulations.

\section*{Usage}
data("PorgMSE")

\section*{Format}

The format is: Formal class 'MSE' [package "DLMtool"] with 17 slots ..@ Name : chr "Stock:Porgy Fleet:Generic_IncE Observation model:Imprecise_Unbiased" ..@ nyears : num 50 ..@ proyears: num 20 ..@ nMPs : int 4 ..@ MPs : chr [1:4] "AvC" "THC" "matlenlim" "area1_50" ..@ nsim : num 20 ..@ OM :'data.frame': 20 obs. of 34 variables: .. ..\$ RefY : num [1:20] 44.8101 .2119 .9 84.3126 ... .. ..\$ M : num [1:20] 0.2420 .2420 .240 .220 .211 ... .. ..\$ Depletion : num [1:20] 0.0791 0.370 .52310 .0870 .3618 ... .. .. \$ A : num [1:20] 53.7659 .31074 .2240 .2 281.7 ... .. ..\$ BMSY_B0 : num [1:20] 0.4290 .4190 .4210 .3940 .414 ... .. ..\$ FMSY_M : num [1:20] 0.2170 .2430 .194 0.4040 .22 ... .. ..\$ Mgrad : num [1:20] \(0.05560 .01920 .03650 .08-0.1671\)... .. .. \$ Msd : num [1:20] 0.083020 .020840 .033840 .034720 .00318 ... .. ..\$ procsd : num [1:20] 0.4920 .310 .4010 .3 0.432 ... .. .. \(\$\) Esd : num [1:20] 0.3910 .1940 .1940 .3660 .38 ... .. .. \(\$\) dFfinal : num [1:20] 0.0276 0.01660 .01560 .02490 .0181 ... .. .. \(\$\) MSY : num [1:20] 58.155 .856 .554 .651 .9 ... .. .. \(\$\) qinc : num [1:20] 0.501-0.656 1.35-1.478 1.554 ... .. ..\$ qcv : num [1:20] 0.1280 .2880 .1630 .1630 .263 ... .. ..\$ FMSY : num [1:20] 0.05260 .0590 .04650 .08890 .0465 ... .. ..\$ Linf : num [1:20] 53.551 50.251 .251 .7 ... .. .. \(\$ \mathrm{~K}\) : num [1:20] 0.1890 .2090 .20 .1860 .19 ... .. .. \$ t0 : num [1:20] -1.02 -1.01-1.04-1.02-1.01 ... .. .. \$ hs : num [1:20] 0.3970 .4080 .3830 .4590 .366 ... .. .. \(\$\) Linfgrad : num [1:20] 0.07810 .14210 .14780 .12720 .0708 ... .. .. \$ Kgrad : num [1:20] 0.0737 -0.1356 \(0.2391-0.0225-0.1415\)... .. ..\$ Linfsd : num [1:20] 0.004640 .014460 .01040 .003580 .01344 ... .. ..\$ recgrad : num [1:20] -6.94-6.77 3.21-8.52 1.53 ... .. ..\$ Ksd : num [1:20] 0.003840 .02322 0.004560 .015380 .01945 ... .. ..\$ ageM : num [1:20] 1.472 .593 .162 .753 .15 ... .. ..\$ V26 : num [1:20] 4.427 .946 .13610 .87 ... .. .. \$ V27 : num [1:20] 2120.32423 .124 .7 ... .. ..\$ V28 : num [1:20] 0.8550 .5360 .9270 .1070 .626 ... .. .. \(\$\) LFC : num [1:20] 12.412 .112 .312 .313 .6 ... .. .. \(\$\) OFLreal : num [1:20] 9.2440 .8544 .7411 .0337 .62 ... .. .. \$ Spat_targ : num [1:20] 11111111 11 ... .. ..\$ Frac_area_1 : num [1:20] 0.10340 .10140 .10380 .09790 .1012 ... .. ..\$ Prob_staying: num [1:20] 0.9760 .9760 .9370 .9860 .926 ... .. .. \$ AC : num [1:20] 0.3220 .6140 .7610 .4780 .31 ... ..@ Obs :'data.frame': 20 obs. of 25 variables: .. ..\$ Cbias : num [1:20] 1.061 .0111 .041 .01 ... .. ..\$ Csd : num [1:20] 0.550 .3990 .2880 .4060 .373 ... .. ..\$ CAA_nsamp : num [1:20] 9253 8370766871957976 ... .. .. \(\$\) CAA_ESS : num [1:20] 14191119141920131816 ... .. .. \(\$\) CAL_nsamp : num [1:20] 9066.686 .858 .265 .1 ... .. ..\$ CAL_ESS : num [1:20] 1517161312 1513181317 ... .. .. \$ Isd : num [1:20] 0.4390 .5630 .5710 .210 .512 ... .. ..\$ Dbias : num [1:20] 1.0380 .9360 .770 .8730 .846 ... .. ..\$ Derr : num [1:20] 0.12930 .15010 .07560 .14220 .0647 ... .. ..\$ Mbias : num [1:20] 0.9351 .0480 .9861 .0091 .028 ... .. ..\$ FMSY_Mbias : num [1:20] 1.121 0.7680 .8811 .0630 .862 ... .. ..\$ BMSY_B0bias: num [1:20] 1.0560 .9780 .7811 .1110 .819 ... .. ..\$ lenMbias : num [1:20] 1.0410 .9391 .0291 .0191 .02 ... .. .. \(\$\) LFCbias : num [1:20] 0.8981 .007 0.9030 .9890 .973 ... .. ..\$ LFSbias : num [1:20] 0.9711 .0131 .0591 .0271 .034 ... .. .. \(\$\) Abias : num [1:20] 1.0451 .0790 .5250 .8262 .767 ... .. .. \$ Aerr : num [1:20] 0.4230 .2110 .4990 .4930 .299 ... .. ..\$ Kbias : num [1:20] 1.0560 .9790 .9391 .011 .019 ... .. .. \$ t0bias : num [1:20] 0.9940 .943 0.9850 .9930 .995 ... .. .. \(\$\) Linfbias : num [1:20] 0.931 .041 .0760 .9440 .994 ... .. ..\$ hbias : num [1:20] 0.9610 .9251 .0281 .0541 .002 ... .. ..\$ Irefbias : num [1:20] 1.1170 .9790 .8691 .0271 .146 ... .. ..\$ Crefbias : num [1:20] 0.9691 .1370 .8731 .0350 .941 ... .. ..\$ Brefbias : num [1:20] 1.015 1.0150 .8711 .0231 .004 ... .. ..\$ betas : num [1:20] 0.9670 .8150 .8131 .3570 .74 ... ..@ B_BMSY : num [1:20, 1:4, 1:20] 0.1130 .6440 .5160 .1710 .581 ... ..@ F_FMSY : num [1:20, 1:4, 1:20] 8.68 1.681 .625 .011 .99 ... ..@ B : num [1:20, 1:4, 1:20] 135783665190804 ... ..@ FM : num [1:20, 1:4, 1:20] 0.45650 .09880 .07520 .44540 .0923 ... ..@ C : num [1:20, 1:4, 1:20] 73.965 .250 .161 .6 64.1 ... ..@ TAC : num [1:20, 1:4, 1:20] 73.965 .250 .161 .664 .1 ... ..@ SSB_hist: num [1:20, 1:34, 1:50, 1:2] 2.372 .281 .181 .481 .64 ... ..@ CB_hist : num [1:20, 1:34, 1:50, 1:2] 0000000000 ... ..@ FM_hist : num [1:20, 1:34, 1:50, 1:2] 0000000000 ...

\section*{Examples}
data(PorgMSE)

Pplot A projection by projection plot of \(F / F M S Y\) and \(B / B M S Y\)

\section*{Description}

A shorter version of the plot method for MSEs that just shows the projected trends in stock status and over exploitation

\section*{Usage}

Pplot(MSEobj, nam=NA)

\section*{Arguments}
\begin{tabular}{ll} 
MSEobj & An object of class MSE \\
nam & Name of the plot
\end{tabular}

Author(s)
T. Carruthers
qopt \begin{tabular}{l} 
Internal optimization function that find the catchability ( \(q\) where \\
\(F=q E\) ) value required to get to user-specified stock depletion (current \\
biomass / unfished biomass)
\end{tabular}

\section*{Description}

The user specifies the level of stock depleiton. This funciton takes the derived effort trajectories and finds the catchabiltiy to get the stock there.

\section*{Usage}
qopt(lnq, depc, Fc, Perrc, Mc, hc, Mac, Wac, R0c, Vc, nyears, maxage, movc, Spat_targc, SRrelc, aRc, bRc, opt=T)

\section*{Arguments}
\(\operatorname{lnq}\)
depc
Fc
Perrc
Mc
hc
Mac
Wac
R0c
Vc
nyears
maxage
move
Spat_targc
SRrelc
aRc
bRc
opt

\section*{Details}

Paired with qopt.

\section*{Author(s)}
T. Carruthers

Rcontrol
Harvest Control Rule using prior for intrinsic rate of increase

\section*{Description}

An MP proposed by Carl Walters that modifies TACs according to trends in apparent surplus production that includes information from a demographically derived prior for intrinsic rate of increase

\section*{Usage}

Rcontrol(x, DLM_data, reps \(=100\), yrsmth \(=10, \mathrm{gg}=2\), glim \(=c(0.5,2)\) )

\section*{Arguments}
\begin{tabular}{ll}
x & A position in data-limited methods data object \\
DLM_data & A data-limited methods data object \\
reps & The number of quota samples \\
yrsmth & The number of years for smoothing catch and biomass data \\
gg & A gain parameters \\
glim & Limits for the change in TAC among years
\end{tabular}

\section*{Author(s)}
C. Walters and T. Carruthers

\section*{References}

Made-up for this package.
Rcontrol2 \begin{tabular}{l} 
MP using prior for intrinsic rate of increase with a quadratic approx- \\
imation to surplus production
\end{tabular}

\section*{Description}

An MP proposed by Carl Walters that modifies quotas according to trends in apparent surplus production that includes information from a demographically derived prior for intrinsic rate of increase. This is different from Rcontrol because it includes a quadratic approximation of recent trend in surplus production given biomass

\section*{Usage}

Rcontrol2(x, DLM_data, reps = 100, yrsmth \(=10\), gg = 2, glim = c(0.5, 2))

\section*{Arguments}
x
DLM_data
reps
yrsmth The number of years for smoothing catch and biomass data
gg A gain parameters
glim Limits for the change in TAC among years

\section*{Author(s)}
C. Walters and T. Carruthers

\section*{References}

Made-up for this package.
replic8 \(\quad\)\begin{tabular}{l} 
Enlarge (replicate) a DLM data object to create an additional dimen- \\
sion for simulation / sensitivity testing
\end{tabular}

\section*{Description}

Replicates position 1 data to multiple positions for sensitivity testing etc

\section*{Usage}
replic8(DLM_data, nrep)

\section*{Arguments}
\begin{tabular}{ll} 
DLM_data & A data-limited methods data object \\
nrep & The number of positions to expand the DLM object to
\end{tabular}

\section*{Author(s)}
T. Carruthers

\section*{Description}

A function that finds all methods in the environment and searches the function text for slots in the DLM data object

\section*{Usage}

Required(funcs = NA)

\section*{Arguments}
funcs A character vector of possible methods of class DLM quota, DLM space or DLM size

\section*{Author(s)}
T. Carruthers
runInMP
Runs input control MPs on a DLM_data object.

\section*{Description}

Function runs a MP (or MPs) of class 'DLM_input' and returns a list: input control recommendation(s) in element 1 and DLM_data object in element 2.

\section*{Usage}
runInMP(DLM_data, MPs = NA, reps = 100)

\section*{Arguments}

DLM_data A object of class DLM_data
MPs A vector of MPs of class 'DLM_input'
reps \(\quad\) Number of stochastic repititions - often not used in input control MPs.

\section*{Author(s)}

\section*{A. Hordyk}
```

runMSE

```

Run a Management Strategy Evaluation

\section*{Description}

A function that runs a Management Strategy Evaluation (closed-loop simulation) for a specified operating model

\section*{Usage}
\[
\begin{aligned}
& \text { runMSE }(O M, \text { MPs }=\text { NA, nsim }=48, \text { proyears }=28, \\
& \quad \text { interval }=4, \text { pstar }=0.5, \operatorname{maxF}=0.8, \\
& \text { timelimit }=1, \text { reps }=1, \text { custompars }=0, \text { CheckMPs=TRUE })
\end{aligned}
\]

\section*{Arguments}
\begin{tabular}{ll} 
OM & An operating model object (class OM) \\
MPs & A vector of methods (character string) of class DLM_output or DLM_input. \\
nsim & Number of simulations \\
proyears & Number of projected years \\
interval & \begin{tabular}{l} 
The assessment interval - how often would you like to update the management \\
system?
\end{tabular}
\end{tabular}
\begin{tabular}{ll} 
pstar & The percentile of the sample of the management recommendation for each method \\
maxF & \begin{tabular}{l} 
Maximum instantaneous fishing mortality rate that may be simulated for any \\
given age class
\end{tabular} \\
timelimit & \begin{tabular}{l} 
Maximum time taken for a method to carry out 10 reps (methods are ignored \\
that take longer)
\end{tabular} \\
reps & \begin{tabular}{l} 
Number of samples of the management recommendation for each method. Note \\
that when this is set to 1, the mean value of the data inputs is used.
\end{tabular} \\
custompars & \begin{tabular}{l} 
A data.table with nsim rows and nparameter columns. The column names must \\
respond to variables of the operating model or observation model see the OM \\
and Obs slots of the MSE class for correct names and interpretation. This allows \\
users to prescribe correlated parameters or estimates from stock assessments.
\end{tabular} \\
CheckMPs & \begin{tabular}{l} 
Logical to indicate if Can function should be used to check if MPs can be run.
\end{tabular}
\end{tabular}

\section*{Value}

An object of class MSE

\section*{Author(s)}
T. Carruthers

\section*{Description}

Run a Management Strategy Evaluation and save out the results to a Rdata file. To increase speed and efficiency, particulary for runs with a large number simulations (nsim), the simulations are split into a number of packets. The functions loops over the packets and combines the output into a single MSE object. If the MSE model crashes during a run, the MSE is run again until it is successfully completed. The MSE is stopped if the number of consecutive crashes exceeds maxCrash. There is an ption to save the packets as Rdata files to the current working directory (default is FALSE). By default, the functions saves the completed MSE object as a Rdata file (to the current working directory).

\section*{Usage}
runMSErobust \((0 M=" 1 ", M P s=N A\), nsim \(=200\), proyears \(=28\), interval \(=4\), pstar \(=0.5\), maxF \(=0.8\), timelimit \(=1\), reps \(=1\), custompars \(=0\), CheckMPs \(=\) TRUE, maxsims \(=64\), name \(=\) NULL, maxCrash \(=10\), saveMSE \(=\) TRUE, savePack \(=\) FALSE)
runMSErobust

\section*{Arguments}
\begin{tabular}{ll} 
OM & An operating model object (class OM) \\
MPs & \begin{tabular}{l} 
A vector of methods (character string) of class DLM_output or DLM_input. If \\
NA all available MPs are run.
\end{tabular} \\
nsim & Number of simulations \\
proyears & Number of projected years \\
interval & \begin{tabular}{l} 
The assessment interval - how often would you like to update the management \\
system?
\end{tabular} \\
pstar & \begin{tabular}{l} 
The percentile of the sample of the management recommendation for each method
\end{tabular} \\
maxF & \begin{tabular}{l} 
Maximum instantaneous fishing mortality rate that may be simulated for any \\
given age class
\end{tabular} \\
timelimit & \begin{tabular}{l} 
Maximum time taken for a method to carry out 10 reps (methods are ignored \\
that take longer)
\end{tabular} \\
reps & \begin{tabular}{l} 
Number of samples of the management recommendation for each method. Note \\
that when this is set to 1, the mean value of the data inputs is used.
\end{tabular} \\
custompars & \begin{tabular}{l} 
A data.table with nsim rows and nparameter columns. The column names must \\
respond to variables of the operating model or observation model see the OM \\
and Obs slots of the MSE class for correct names and interpretation. This allows
\end{tabular} \\
users to prescribe correlated parameters or estimates from stock assessments.
\end{tabular}

\section*{Value}

An object of class MSE

\section*{Author(s)}
A. Hordyk and T. Carruthers

\section*{Description}

A wrapper function that gets the OFL recommendation in cases where a method of DLM quota has been specified

\section*{Usage}

Sam(DLM_data, MPs = NA, reps = 100, maxlines = 10, perc = 0.5)

\section*{Arguments}
\begin{tabular}{ll} 
DLM_data & A data-limited methods data object \\
MPs & A character vector of methods of DLM quota, DLM space or DLM size \\
reps & The number of samples of quota recommendations by method \\
maxlines & \\
perc &
\end{tabular}

\section*{Author(s)}
T. Carruthers

\section*{Description}

An MP that makes incremental adjustments to TAC recommendations based on the apparent trend in CPUE

\section*{Usage}

SBT1 (x, DLM_data, reps = 100, yrsmth=10, k1=1.5, k2=3, gamma=1)

\section*{Arguments}
\(x \quad\) A position in a data-limited methods data object
DLM_data A data-limited methods data object
reps The number of samples of the TAC
yrsmth The number of years for evaluating trend in relative abundance indices
k1 Control parameter
k2 Control parameter
gamma Control parameter

\section*{Details}

This isn't exactly the same as the proposed methods and is stochastic in this implementation. The method doesn't tend to work too well under many circumstances possibly due to the lack of 'tuning' that occurs in the real SBT assessment environment. You could try asking Rich Hillary at CSIRO about this approach.

\section*{Author(s)}
T. Carruthers

\section*{References}
http://www.ccsbt.org/site/recent_assessment.php
SBT2 SBT complex MP

\section*{Description}

An MP that makes incremental adjustments to TAC recommendations based on index levels relative to target levels (BMSY/B0) and catch levels relative to target levels (MSY)

\section*{Usage}

SBT2 (x, DLM_data, reps \(=100\), epsB=0.25,epsR=0.75,tauR=5,tauB=7,gamma=1)

\section*{Arguments}
\begin{tabular}{ll}
\(x\) & A position in a data-limited methods data object \\
DLM_data & A data-limited methods data object \\
reps & The number of samples of the TAC \\
epsB & Control parameter \\
epsR & Control parameter \\
tauR & Control parameter \\
tauB & Control parameter \\
gamma & Control parameter
\end{tabular}

\section*{Details}

This isn't exactly the same as the proposed methods and is stochastic in this implementation. The method doesn't tend to work too well under many circumstances possibly due to the lack of 'tuning' that occurs in the real SBT assessment environment. You could try asking Rich Hillary at CSIRO about this approach.

\section*{Author(s)}
T. Carruthers

\section*{References}
http://www.ccsbt.org/site/recent_assessment.php
\begin{tabular}{ll}
\hline Sense \(\quad\) Sensitivity analysis \\
\hline
\end{tabular}

\section*{Description}

A function that determines the inputs for a given data-limited method of class DLM_output and then analyses the sensitivity of TAC estimates to marginal differences in each input. The range used for sensitivity is based on the user-specified CV for that input (e.g. CV_Mort, Mort)

\section*{Usage}

Sense(DLM_data, MP, nsense \(=6\), reps \(=100\), perc \(=c(0.05,0.5,0.95)\), ploty \(=T)\)

\section*{Arguments}

DLM_data A data-limited methods data object (class DLM_data)
MP A character string representing an MP applied in calculating the TAC recommendations in the DLM object
nsense \(\quad\) The number of points over which to calculate the TAC (resolution)
reps \(\quad\) The number of samples of the quota taken for the calculation of the TAC
perc \(\quad\) The percentile of the sample TAC
ploty A logical switch, (T/F, should a plot be drawn?)

\section*{Author(s)}
T. Carruthers

SetRecruitCycle Function to calculate cyclic recruitment pattern given user-specified values of period and amplitude.

\section*{Description}

Calculates cyclic pattern in recruitment deviations for a simulation. Ranges for Period and Amplitude are specified by user, and function produces cyclic pattern from within these ranges. Default is a sine wave.

\section*{Usage}

SetRecruitCycle(x=1, Period, Amplitude, TotYears, Shape=c("sin", "shift"))

\section*{Arguments}
\begin{tabular}{ll}
x & Simulation number. \\
Period & \begin{tabular}{l} 
A vector of length 2 specifying the minimum and maximum values for the pe- \\
riod of the recruitment cycles. e.g., if Period \(=\mathrm{c}(10,10)\), then recruitment cycle \\
occurs every 10 years exactly.
\end{tabular} \\
Amplitude & \begin{tabular}{l} 
A vector of length 2 specifying the minimum and maximum values for the am- \\
plitude of the recruitment cycles. e.g., if Amplitude \(=\mathrm{c}(0,0.5)\), the average \\
recruitment will increase (or decrease) by a factor between 0 and 0.5 each cycle.
\end{tabular} \\
TotYears & \begin{tabular}{l} 
A numeric value specifying the total number of years (should be nyears + proyears). \\
Shape
\end{tabular} \begin{tabular}{l} 
Specifies whether cyclic recruitment pattern is sine wave (default) or a step- \\
change (shift).
\end{tabular}
\end{tabular}

\section*{Author(s)}
A. Hordyk
slotlim An data-limited method which sets a slot limit

\section*{Description}

An example of the implementation of input controls in the DLM toolkit, where selectivity-at-length is set using a slot limit; that is, a minimum and maximum legal length. The maximum limit is set here, quite arbitrarily, as the 75th percentile between the new minimum legal length and the estimated asymptotic length.

\section*{Usage}
slotlim(x, DLM_data, ...)

\section*{Arguments}
\begin{tabular}{ll}
x & A position in a data-limited methods object \\
DLM_data & A data-limited methods object \\
\(\ldots\) & \begin{tabular}{l} 
Optional additional arguments that are ignored. Note arguments reps or \(\ldots\) \\
are required for all input controls
\end{tabular}
\end{tabular}

\section*{Value}

A vector of input control recommendations, with values for length at first capture, full selection, and maximum size limit in the 5th, 6th, and 7th elements of the vector

\section*{Author(s)}
A. Hordyk

\section*{References}

Made-up for this package

SnapMSE Example MSE object used in the vignette

\section*{Description}

A dummy example MSE object, with blue shark, generic fleet and imprecise and biased observation model, four MPs, and 16 simulations.

\section*{Usage}
data("SnapMSE")

\section*{Format}

The format is: Formal class 'MSE' [package "DLMtool"] with 17 slots ..@ Name : chr "Stock:Blue_shark Fleet:Generic_fleet Observation model:Imprecise_Biased" ..@ nyears : num 50 ..@ proyears: num 30 ..@ nMPs : int 4 ..@ MPs : chr [1:4] "Fratio" "DCAC" "Fdem" "DD" ..@ nsim : num 16 ..@ OM :'data.frame': 16 obs. of 34 variables: .. ..\$ RefY : num [1:16] 1591674011048028878534 ... .. ..\$ M : num [1:16] 0.1720 .1750 .1670 .2420 .185 ... .. ..\$ Depletion : num [1:16] 0.3970 .555 0.3610 .3380 .57 ... .. .. \$ A : num [1:16] 631441191347161184138 ... .. ..\$ BMSY_B0 : num [1:16] 0.380 .3460 .3040 .3370 .349 ... .. ..\$ FMSY_M : num [1:16] 0.4970 .5770 .830 .4230 .478 ... .. ..\$ Mgrad : num [1:16] -0.1167-0.2097-0.0849 \(0.197-0.0967\)... .. ..\$ Msd : num [1:16] 0.064110 .080610 .091910 .005860 .04776 ... .. .. \$ procsd : num [1:16] 0.2480 .2130 .1520 .233 0.25 ... .. ..\$ Esd : num [1:16] 0.3150 .3980 .2140 .3330 .38 ... .. ..\$ dFfinal : num [1:16] 0.00832 -0.00412 \(0.004920 .01095-0.00757\)... .. .. \(\$\) MSY : num [1:16] 25733845467428844262 ... .. .. \$ qinc : num [1:16] -0.331 0.525-1.985 0.710 .915 ... .. ..\$ qcv : num [1:16] 0.1810 .2180 .151 0.2510 .133 ... .. ..\$ FMSY : num [1:16] 0.08550 .10110 .13870 .10250 .0882 ... .. ..\$ Linf : num [1:16] 197196201197201 ... .. .. \$ K : num [1:16] 0.2260 .2320 .2390 .2380 .218 ... .. .. \(\$\) t0 :
num [1:16] -1.032-1.027-0.971-1.011-0.959 ... .. .. \$ hs : num [1:16] 0.4870 .6470 .7690 .656 0.638 ... .. ..\$ Linfgrad : num [1:16] 0.1212-0.0666-0.1746-0.1365 0.1226 ... .. ..\$ Kgrad : num [1:16] \(0.01160 .1432-0.2061-0.1219-0.0717\)... .. .. \(\$\) Linfsd : num [1:16] 0.00440 .015070 .02219 0.012260 .00046 ... .. .. \$ recgrad : num [1:16] -8.58 1.41 1.21-1.32 9.24 ... .. ..\$ Ksd : num [1:16] 0.02060 .02450 .01590 .01730 .0198 ... .. .. \(\$\) ageM : num [1:16] 4.373 .813 .84 .334 .12 ... .. .. \(\$\) V26 : num [1:16] 35.74631 .133 .444 ... .. ..\$ V27 : num [1:16] 147103117108117 ... .. ..\$ V28 : num [1:16] 0.1640 .4560 .2970 .7970 .695 ... .. .. \$ LFC : num [1:16] 55.948 .96052 .258 ... .. .. \$ OFLreal : num [1:16] 20354356583734274114 ... .. .. \(\$\) Spat_targ : num [1:16] 11111111 11 ... .. ..\$ Frac_area_1 : num [1:16] 0.09520 .10280 .09830 .09970 .1023 ... .. ..\$ Prob_staying: num [1:16] 0.850 .8330 .8790 .8780 .829 ... .. ..\$ AC : num [1:16] 0.120 .7050 .6110 .1460 .639 ... ..@ Obs :'data.frame': 16 obs. of 25 variables: .. ..\$ Cbias : num [1:16] 0.6041 .0990 .7871 .049 0.999 ... .. ..\$ Csd : num [1:16] 0.3750 .2890 .4290 .5120 .262 ... .. ..\$ CAA_nsamp : num [1:16] 89808665738897547057 ... .. ..\$ CAA_ESS : num [1:16] 14131713152015121519 ... .. ..\$ CAL_nsamp : num [1:16] 77.370 .678 .596 .873 .9 ... .. ..\$ CAL_ESS : num [1:16] 2011 1719161620141519 ... .. ..\$ Isd : num [1:16] 0.2740 .3410 .470 .5280 .342 ... .. .. \$ Dbias : num [1:16] 0.7621 .4241 .640 .6970 .599 ... .. ..\$ Derr : num [1:16] 0.07470 .08840 .09180 .0827 0.0928 ... .. .. \$ Mbias : num [1:16] 0.7490 .7272 .7110 .4370 .981 ... .. .. \$ FMSY_Mbias : num [1:16] 0.8621 .3360 .2731 .2680 .683 ... .. ..\$ BMSY_B0bias: num [1:16] 0.6850 .9411 .1921 .006 1.446 ... .. ..\$ lenMbias : num [1:16] 0.8280 .7341 .0330 .6720 .816 ... .. ..\$ LFCbias : num [1:16] 0.9220 .881 1.1 0.9931 .093 ... .. .. \(\$\) LFSbias : num [1:16] 0.8840 .9580 .8611 .041 ... .. .. \(\$\) Abias : num [1:16] 0.5383 .3033 .9030 .9690 .975 ... .. .. \(\$\) Aerr : num [1:16] 0.3110 .2880 .3930 .3170 .21 ... .. ..\$ Kbias : num [1:16] 1.2650 .9051 .0520 .9740 .899 ... .. ..\$ t0bias : num [1:16] 1.0430 .956 1.0910 .9941 .103 ... .. .. Linfbias : num [1:16] 0.8050 .9750 .9640 .8951 .037 ... .. .. \(\$\) hbias : num [1:16] 0.890 .9450 .9211 .191 .007 ... .. .. \(\$\) Irefbias : num [1:16] 1.9280 .8360 .8450 .5810 .706 ... .. .. \(\$\) Crefbias : num [1:16] 0.8761 .0520 .7920 .5530 .672 ... .. .. \$ Brefbias : num [1:16] 0.3340 .458 1.1941 .0031 .656 ... .. .. \(\$\) betas : num [1:16] 2.9882 .7510 .4532 .7272 .308 ... ..@ B_BMSY : num [1:16, 1:4, 1:30] 0.6381 .171 .7581 .3590 .861 ... ..@ F_FMSY : num [1:16, 1:4, 1:30] 0.373 2.8541 .1980 .6680 .83 ... ..@ B : num [1:16, 1:4, 1:30] 3401767403976384428649632 ... ..@ FM : num [1:16, 1:4, 1:30] 0.03190 .28860 .16610 .06850 .0732 ... ..@ C : num [1:16, 1:4, 1:30] \(649156471147328523300 \ldots\)..@ TAC : num [1:16, 1:4, 1:30] \(649156471147328523300 \ldots\) ..@ SSB_hist: num \([1: 16,1: 46,1: 50,1: 2] 3.1442 .7612 .8319 .0550 .222\)... ..@ CB_hist : num [1:16, 1:46, 1:50, 1:2] 0000000000 ... ..@ FM_hist : num [1:16, 1:46, 1:50, 1:2] 0000000 000 ...

\section*{Examples}
```

data(SnapMSE)

```

\section*{SPmod} Surplus production based catch-limit modifier

\section*{Description}

An MP that makes incremental adjustments to TAC recommendations based on the apparent trend in surplus production. Based on the theory of Mark Maunder (IATTC)

\section*{Usage}

SPmod(x, DLM_data, reps \(=100\), alp \(=c(0.8,1.2)\), bet \(=c(0.8,1.2))\)

\section*{Arguments}
x
DLM_data A data-limited methods data object
reps The number of quota samples
alp Condition for modifying the TAC (bounds on change in abundance)
bet Limits for how much the TAC can change among years

\section*{Details}

Note that this isn't exactly what Mark has previously suggested and is stochastic in this implementation.

\section*{Value}

A numeric vector of TAC recommendations

\section*{Author(s)}
T. Carruthers

\section*{References}
http://www.iattc.org/Meetings/Meetings2014/MAYSAC/PDFs/SAC-05-10b-Management-StrategyEvaluation.pdf

\section*{Description}

An MP that uses Martell and Froese (2012) method for estimating MSY to determine the OFL. Since their approach estimates stock trajectories based on catches and a rule for intrinsic rate of increase it also returns depletion. Given their surplus production model predicts K , r and depletion it is straighforward to calculate the OFL based on the Schaefer productivity curve. OFL = dep \(x\) (1-dep) x r x K x 2

\section*{Usage}

SPMSY(x, DLM_data, reps = 100)

\section*{Arguments}
\(x \quad\) A position in a data-limited methods data object
DLM_data A data-limited methods data object
reps \(\quad\) The number of samples of the TAC

\section*{Details}

Requires the assumption that catch is proportional to abundance. Occasionally the rule that limits r and K ranges does not allow \(\mathrm{r}-\mathrm{K}\) pairs to be found that lead to the depletion inferred by the catch trajectories. In this case this method widens the search.

\section*{Author(s)}
T. Carruthers

\section*{References}

Martell, S. and Froese, R. 2012. A simple method for estimating MSY from catch and resilience. Fish and Fisheries. DOI: 10.1111/j.1467-2979.2012.00485.x
```

SPslope Slope in surplus production MP

```

\section*{Description}

A management procedure that makes incremental adjustments to TAC recommendations based on the apparent trend in recent surplus production. Based on the theory of Mark Maunder (IATTC)

\section*{Usage}

SPslope (x, DLM_data, reps \(=100\), yrsmth \(=4\), alp \(=c(0.9,1.1)\), bet \(=c(1.5,0.9))\)

\section*{Arguments}
x
DLM_data A data-limited methods data object
reps The number of quota samples
yrsmth Years over which to smooth recent estimates of surplus production
alp Condition for modifying the DLM_data (bounds on change in abundance)
bet Limits for how much the DLM_data can change among years

\section*{Details}

Note that this isn't exactly what Mark has previously suggested and is stochastic in this implementation.

\section*{Value}

A numeric vector of DLM_data recommendations

\section*{Author(s)}
T. Carruthers

\section*{References}
http://www.iattc.org/Meetings/Meetings2014/MAYSAC/PDFs/SAC-05-10b-Management-StrategyEvaluation.pdf

SPSRA Surplus Production Stock Reduction Analysis

\section*{Description}

A surplus production equivalent of DB-SRA that uses a demographically derived prior for intrinsic rate of increase (McAllister method, below)

\section*{Usage}

SPSRA(x, DLM_data, reps = 100)

\section*{Arguments}
\(x \quad\) A position in a data-limited methods data object
DLM_data A data-limited methods data object (class DLM)
reps \(\quad\) The number of samples of the TAC taken for the calculation of the quota

\section*{Author(s)}
T. Carruthers

\section*{References}

McAllister, M.K., Pikitch, E.K., and Babcock, E.A. 2001. Using demographic methods to construct Bayesian priors for the intrinsic rate of increase in the Schaefer model and implications for stock rebuilding. Can. J. Fish. Aquat. Sci. 58: 1871-1890.
```

SPSRA_ML Surplus Production Stock Reduction Analysis using a mean-length es-
timate of current stock depletion

```

\section*{Description}

A surplus production equivalent of DB-SRA that uses a demographically derived prior for intrinsic rate of increase. A prior for depletion is calculated from a mean-length estimator

\section*{Usage}

SPSRA_ML(x, DLM_data, reps = 100)

\section*{Arguments}
x
DLM_data A data-limited methods data object (class DLM)
reps \(\quad\) The number of samples of the TAC taken

\section*{Note}

The mean length extension was programmed by Gary Nelson as part of his excellent R package 'fishmethods'

\section*{Author(s)}
T. Carruthers

\section*{References}

McAllister, M.K., Pikitch, E.K., and Babcock, E.A. 2001. Using demographic methods to construct Bayesian priors for the intrinsic rate of increase in the Schaefer model and implications for stock rebuilding. Can. J. Fish. Aquat. Sci. 58: 1871-1890.
```

Stock-class Class "Stock"

```

\section*{Description}

An operating model component that specifies the parameters of the population dynamics model

\section*{Objects from the Class}

Objects can be created by calls of the form new("Stock", OM).

\section*{Slots}

Name: The name of the Stock object
maxage: The maximum age of individuals that is simulated (there is no 'plus group': individuals die off beyone the maximum age so there isn't a huge cost to simulating more older age classes)
\(R 0\) : The magnitude of unfished recruitment. This is normally fixed to some arbitrary value since it simply scales the simulated numbers)
M: Natural mortality rate (uniform distribution)
Msd: Inter-annual variability in natural mortality rate expressed as a coefficient of variation (uniform distribution)
Mgrad: Mean temporal trend in natural mortality rate, expressed as a percentage change in \(M\) per year (uniform distribution)
h: Steepness of the stock recruit relationship (uniform distribution)
SRrel: Type of stock-recruit relationship (1)Beverton-Holt (2) Ricker
Linf: Maximum length (uniform distribution)
\(K\) : von \(B\). growth parameter \(k\) (uniform distribution)
t0: von B. theoretical age at length zero (uniform distribution)
Ksd: Inter-annual variability in growth parameter k (uniform distribution)
Kgrad: Mean temporal trend in growth parameter k , expressed as a percentage change in k per year (uniform distribution)
Linfsd: Inter-annual variability in maximum length - uniform distribution
Linfgrad: Mean temporal trend in maximum length, expressed as a percentage change in Linf per year (uniform distribution)
recgrad: Mean temporal trend in log-normal recruitment deviations (uniform distribution)
AC : Autocorrelation in recruitment deviations \(\operatorname{rec}(\mathrm{t})=\mathrm{AC} * \operatorname{rec}(\mathrm{t}-1)+(1-\mathrm{AC}) * \operatorname{sigma}(\mathrm{t})\) (uniform distribution)
a: Length-weight parameter alpha (uniform distribution)
b: Length-weight parameter beta (uniform distribution)
L50: Length-at- 50 percent maturity (uniform distribution)
L50_95: Length increment from 50 percent to 95 percent maturity
D: Current level of stock depletion (Bcurrent/Bunfished) (uniform distribution)
Perr: Process error, the CV of lognormal recruitment deviations (uniform distribution)
Period: Period for cylical recruitment pattern in years (uniform distribution). Leave empty to ignore
Amplitude: Amplitude in deviation from long-term average recruitment during recruitment cycle, both positive and negative (uniform distribution). E.g., a range from 0 to 0.5 means recruitment decreases or increases by up to \(50 \%\) each cycle. Leave empty to ignore
Size_area_1: The size of area 1 relative to area 2 (uniform distribution)
Frac_area_1: The fraction of the unfished biomass in stock 1 (uniform distribution)
Prob_staying: The probability of inviduals in area 1 remaining in area 1 over the course of one year
Source: A reference to a website or article form which parameters were taken to define the operating model

\section*{Methods}
initialize signature(.Object = "Stock"): ...

\section*{Author(s)}
T. Carruthers

\section*{Examples}
showClass("Stock")

\section*{Description}

Subset the MSE object by particular MPs (either MP number or name), or particular simulations, or a subset of the projection years (e.g., 1: < projection years).

\section*{Usage}

Sub(MSEobj, MPs=NULL, sims=NULL, years=NULL)

\section*{Arguments}

MSEobj
MPs A vector MPs names or MP numbers to subset the MSE object. Defaults to all MPs.
sims A vector of simulation numbers to subset the MSE object. Can also be a logical vector. Defaults to all simulations.
years A numeric vector of projection years. Should start at 1 and increase by one to some value equal or less than the total number of projection years.

\section*{Author(s)}
A. Hordyk

\section*{Description}
\(\sim \sim\) Methods for function summary \(\sim \sim\)

\section*{Methods}
signature(object = "DLM")
signature(object = "MSE")

\section*{Description}

A function that returns the stochastic TAC recommendations from a vector of data-limited MPs (DLM_output) given a data-limited data object DLM_data

\section*{Usage}

TAC(DLM_data, MPs = NA, reps = 100, maxlines = 6, perc = NA, xlims = NA, timelimit = 1)

\section*{Arguments}

DLM_data
MPs
reps
maxlines
perc
xlims
timelimit

\section*{Author(s)}
T. Carruthers

\section*{Tplot A trade-off plot for an MSE object}

\section*{Description}

A shorter version of the plot method for MSEs that just shows the overall trade-offs

\section*{Usage}

Tplot(MSEobj, nam=NA)

\section*{Arguments}
\begin{tabular}{ll} 
MSEobj & An object of class 'MSE' \\
nam & Name of the plot
\end{tabular}

\section*{Author(s)}
T. Carruthers
```

Tplot2

```

A trade-off plot for an MSE object that compares long-term yield (LTY: fraction of simulations getting over half FMSY yield in the last ten years of the projection), short-term yield (STY: fraction of simulations getting over half FMSY yield in the first ten years of the projection), variability in yield (VY: fraction of simulations where average annual variability in yield is less than 10 per cent) and biomass level (B10: the fraction of simulations in which biomass stays above 10 percent of BMSY).

\section*{Description}

A shorter version of the plot method for MSEs that just shows the overall trade-offs

\section*{Usage}

Tplot2(MSEobj, nam=NA)

\section*{Arguments}

MSEobj An object of class 'MSE'
nam Name of the plot

\section*{Author(s)}
T. Carruthers

\section*{Description}

Creates a trade-off plot (up to four panels) of built-in performance metrics.

\section*{Usage}

TradePlot(MSEobj, XAxis=c("Overfishing", "Biomass:BMSY"),
YAxis=c("Long-term Yield", "AnnualVar"), XThresh=c(30, 80), YThresh=c(0,50), maxVar=15, BmsyRef=0.5, B0Ref=0.2, AvailMPs=NULL, ShowLabs=FALSE, ShowCols=TRUE)

\section*{Arguments}
\begin{tabular}{ll} 
MSEobj & Object of class MSE, output of the runMSE function \\
XAxis & \begin{tabular}{l} 
Character string describing the performance metrics for the x-axis (or x-axes if \\
vector; max 4). Must be chosen for list of existing PMs and same length as \\
YAxis. See PMs
\end{tabular} \\
YAxis & \begin{tabular}{l} 
Character string describing the performance metrics for the y-axis (or y-axes if \\
vector; max 4). Must be chosen for list of existing PMs and same length as \\
XAxis. See PMs
\end{tabular} \\
XThresh & \begin{tabular}{l} 
Minimum threshold values in percent (i.e., \(50=50 \%\) ) for the x-axes (must be \\
same length as XAxis)
\end{tabular} \\
YThresh & \begin{tabular}{l} 
Minimum threshold values in percent (i.e., \(50=50 \%\) ) for the y-axes (must be \\
same length as YAxis)
\end{tabular} \\
maxVar & \begin{tabular}{l} 
Reference for average annual variability in yield in percent
\end{tabular} \\
BmsyRef & \begin{tabular}{l} 
Reference level of BMSY, in proportion, i.e., \(0.5=0.5 B M S Y\) \\
B0Ref
\end{tabular} \\
Reference level of B0, in proportion, i.e., \(0.2=0.2 B 0\)
\end{tabular}

\section*{Details}

Returns a list containing the names of performance metrics that meet the minimum performance metrics for each trade-off, and ranks the MPs by increasing distance from the top-right corner.

\section*{Author(s)}
A. Hordyk

\section*{Description}

A function that relates operating model parameters and parameters of the observation model to yield (by default). A user can also specific their own utility values (Ut) which is arranged in a matrix of nsim rows and nMP columns.

\section*{Usage}

VOI(MSEobj, ncomp \(=6\), nbins \(=8\), maxrow \(=8\), Ut \(=\) NA, Utnam = "Utility")

\section*{Arguments}
\begin{tabular}{ll} 
MSEobj & An object of class MSE \\
ncomp & Maximum number of variables to examine per MP \\
nbins & \begin{tabular}{l} 
Number of percentile bins for sampled parameters of the operating model or \\
observation model, which is used for calculating variability in utility across the \\
sampled range of each parameter
\end{tabular} \\
maxrow & maximum number of MPs per plot \\
Ut & A matrix of user-specified utility values of nsim rows and nMPs columns \\
Utnam & The name of the utility measure for plotting
\end{tabular}

\section*{Author(s)}
T. Carruthers

\section*{Description}

A function that relates operating model parameters and parameters of the observation model to relative yield (yield over last 5 years of projection relative to a 'best F ' scenario that maximizes yield).

\section*{Usage}

VOI2(MSEobj, ncomp = 6, nbins = 4, Ut = NA, Utnam = "yield", lay = F)

\section*{Arguments}
\begin{tabular}{ll} 
MSEobj & An object of class MSE \\
ncomp & Maximum number of observation variables to examine per MP \\
nbins & \begin{tabular}{l} 
Number of bins for sampled observation variables used for calculating variabil- \\
ity in utility across the sampled range of each parameter
\end{tabular} \\
Ut & A matrix of user-specified utility values of nsim rows and nMPs columns \\
Utnam & The name of the utility measure for plotting \\
lay & Controls whether labels are in lay terms or not
\end{tabular}

\section*{Note}

VOI2 assumes that relative cost for each type of improvement in data is linearly related to the number of samples (e.g. nCAAobs) or square function of improved precision and bias e.g.: relative cost \(=1 /(\text { newCV/oldCV) })^{\wedge} 2\)

\section*{Author(s)}
T. Carruthers
```

VOIplot Yet another Value of Information Plot

```

\section*{Description}

A function that relates parameters of the observation model and the operating model parameters to yield.

\section*{Usage}

VOIplot(MSEobj, MPs=NA, nvars=5, nMP=4, Par=c("Obs", "OM"), YVar=c("Y", "B"), doPlot=TRUE, incStat=FALSE, availMP=NULL, acceptMP=NULL, incNames=TRUE, labcex=0.8)

\section*{Arguments}
\begin{tabular}{ll} 
MSEobj & An object of class MSE \\
MPs & \begin{tabular}{l} 
The MPs to plot. If NA it will plot the first nMP from MSEobj \\
nvars
\end{tabular} \\
\begin{tabular}{l} 
The number of observation or operating model parameters to plot (number of \\
columns)
\end{tabular} \\
nMP & \begin{tabular}{l} 
The maximum number of MPs to plot (number of rows)
\end{tabular} \\
YVar & Plot Operating Model (OM) or Observation (Obs) parameters? \\
doPlot & Variable for Y-Axis: Yield (Y) or Biomass (B) (relative to BMSY) \\
incStat & Output the plot?
\end{tabular}
\begin{tabular}{ll} 
availMP & \begin{tabular}{l} 
Optional character string of MPs that are available. These names are colored \\
black
\end{tabular} \\
acceptMP & \begin{tabular}{l} 
Optional character string of MPs that are acceptable. These names are colored \\
green if they are also in availMP
\end{tabular} \\
incNames & \begin{tabular}{l} 
Include the names? \\
labcex
\end{tabular} \\
& Character size of the label
\end{tabular}

\section*{Value}

A list of all the information included in the plot

\section*{Author(s)}
A. Hordyk

VPA Robust Virtual Population Analysis

\section*{Description}

A simple 2 parameter ( q , terminal F ) VPA. Note that this is an early version that needs more testing.

\section*{Usage}

VPA(x, DLM_data, reps = reps)

\section*{Arguments}
\(x \quad\) A position in a data-limited methods data object
DLM_data A data-limited methods data object
reps The number of stochastic samples of the TAC recommendation

\section*{Value}

A numeric vector of TAC recommendations

\section*{Note}

This is an early version of the VPA and seems not to be working correctly at present. More testing and development is needed.

\section*{Author(s)}
C. Walters (Model), R. Licandeo (R code), T. Carruthers (DLMtool implementation)

\section*{References}

Method based on VPA of Carl Walters and Roberto Licandeo.
wormplot Biomass wormplot.

\section*{Description}

A worm plot for plotting the likelihood of meeting biomass targets in future years.

\section*{Usage}
wormplot(MSEobj, Bref=0.5, \(\mathrm{LB}=0.25\), UB=0.75)

\section*{Arguments}

MSEobj Object of class MSE, output of the runMSE function
Bref The reference fraction of BMSY (to evaluate the probability of exceeding this level)

LB The lower bound probability that seperates red (bad) and yellow (O.K.) colored segments
UB The upper bound probability that seperates yellow (O.K.) and green (good) colored segments

\section*{Details}

Returns a matrix of nMPs rows and proyears columns which is the fraction of simulations for which biomass was above Bref.

\section*{Author(s)}
T. Carruthers
```

writeCSV

```

Internal function to write CSVs for objects

\section*{Description}

Used internally in the DLMtool package to write CSV files from an existing DLMtool object
```

Usage
writeCSV (inobj, tmpfile = NULL, objtype = c("Stock", "Fleet", "Observation", "DLM_data", "OM", "DLM_fease"))

```

\section*{Arguments}
\begin{tabular}{ll} 
inobj & A object of class Stock, Fleet, Observation, DLM_data, OM, or DLM_fease \\
tmpfile & The full file path and name for the saved CSV file \\
objtype & The class corresonding to the inobj
\end{tabular}

\section*{Author(s)}
A. Hordyk

\section*{Description}

A simple yield per recruit approximation to FMSY (F01) which is the position of the ascending YPR curve for which \(\mathrm{dYPR} / \mathrm{dF}=0.1(\mathrm{dYPR} / \mathrm{d} 0)\)

\section*{Usage}

YPR(x, DLM_data, reps = 100)

\section*{Arguments}
x
A position in a data-limited methods data object
DLM_data
reps

A data-limited methods data object
The number of samples of the TAC

\section*{Value}

A numeric vector of TAC samples

\section*{Note}

Based on the code of Meaghan Bryan

\section*{Author(s)}

Meaghan Bryan and Tom Carruthers

\section*{References}

Beverton and Holt. 1954.

Yield Per Recruit analysis to get FMSY proxy F01 paired to a naive catch curve estimate of recent \(Z\)

\section*{Description}

A simple yield per recruit approximation to FMSY (F01) which is the position of the ascending YPR curve for which \(\mathrm{dYPR} / \mathrm{dF}=0.1(\mathrm{dYPR} / \mathrm{d} 0)\) A naive catch-curve analysis is used to determine recent \(Z\) which given M (Mort) gives F and thus abundance \(=\mathrm{Ct} /(1-\exp (-\mathrm{F}))\)

\section*{Usage}

YPR_CC(x, DLM_data, reps \(=100\), Fmin=0.005)

\section*{Arguments}

X
DLM_data
reps
Fmin

A position in a data-limited methods data object
A data-limited methods data object (class DLM)
The number of samples of the TAC
The minimum fishing mortality rate inferred from the catch-curve analysis

\section*{Author(s)}

Meaghan Bryan and T. Carruthers
\begin{tabular}{ll} 
YPR_ML & \begin{tabular}{l} 
Yield Per Recruit analysis to get FMSY proxy F01 paired with a mean- \\
length estimate of current stock size
\end{tabular}
\end{tabular}

\section*{Description}

A simple yield per recruit approximation to FMSY (F01) which is the position of the ascending YPR curve for which \(\mathrm{dYPR} / \mathrm{dF}=0.1(\mathrm{dYPR} / \mathrm{d} 0)\) A mean-length estimate of recent Z is used to infer current abundance

\section*{Usage}

YPR_ML(x, DLM_data, reps = 100)

\section*{Arguments}
x
DLM_data
reps \(\quad\) The number of samples of the TAC

\section*{Note}

The mean length extension was programmed by Gary Nelson as part of his excellent R package 'fishmethods'

\section*{Author(s)}

Meaghan Bryan and T. Carruthers

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