Package 'FSA'

October 10, 2015
Title Functions for Simple Fisheries Stock Assessment Methods
Version 0.8.1
Date 2015-10-08
Description Functions to perform a variety of simple fish stock assessment methods. Detailed vignettes are available on the fishR website listed below.
<pre>URL http://derekogle.com/fishR/, https://github.com/droglenc/FSA</pre>
<pre>BugReports https://github.com/droglenc/FSA/issues</pre>
License GPL (>= 2)
LazyData true
Depends R (>= 3.0.0)
Imports graphics, grDevices, stats, tools, utils, plotrix, plyr
Suggests car, dplyr, dunn.test, fishmethods, FSAdata, gdata, gplots, Hmisc, knitr, lmtest, marked, nlstools, Rcapture, relax, sciplot, testthat
NeedsCompilation no
Author Derek Ogle [aut, cre]
Maintainer Derek Ogle <derek@derekogle.com></derek@derekogle.com>
Repository CRAN
Date/Publication 2015-10-10 23:04:09
R topics documented:
addRadCap addZeroCatch ageBias ageKey ageKeyPlot 10 agePrecision 18 alkAgeDist 2 alkMeanVar 2

ocFuns	 	 			 		 	 					. 25
oinCI	 	 			 		 	 					. 27
BluegillJL	 	 			 		 	 					. 29
oootstrap	 	 			 		 	 					. 30
BrookTroutTH .	 	 			 		 	 					. 32
capHistConvert .													
capHistSum													
catchCurve													
chapmanRobson													
ChinookArg													
chooseColors													
CodNorwegian													
col2rgbt													
compIntercepts													
compSlopes													
confint.nlsBoot													
CutthroatAL													
depletion													
dunnTest													
Ecoli													
expandCounts													
extraTests													
fact2num													
fishR													
fitPlot													
FSA													
fsaNews													
FSAUtils													
gConvert													
growthModels													
neadtail													
nist.formula													
noCoef													
nyperCI													
													. 97
«Counts													. 101
ksTest													
agratio													
encat													
enFreqExpand .													
logbtcf													
wCompPreds													
mapvalues													
Mirex													
Mmethods													
mrClosed													
oddeven													
perc	 	 			 		 	 					. 130

addRadCap 3

addRa	adCap	dd a total radius-at-capture varial ne-fish-per-line increments data.	ble to a data.frame that contains
Index			191
	ws rai		
	•		
	•		
	psdPlot		
	PSDlit		
	*		
	*		
	•		
	PikeNY		131

Description

Add a total radius-at-capture variable to a data.frame that contains one-fish-per-line increments data (i.e., "wide" format).

Usage

```
addRadCap(df, in.pre = NULL, in.var = NULL, var.name = "radcap")
```

4 addRadCap

Arguments

df	A data frame that contains the growth increment data in one-fish-per-line (i.e., "wide") format.
in.pre	A string that represents the common part of the increment variable names. See details.
in.var	A vector of variables in df that do not change. See details.
var.name	A string that indicates the name for the new total radius-at-capture variable in the new data.frame.

Details

#' The columns that contain the original measurement data can specified in a variety of ways. First, if all columns begin with the same prefix (and no other columns contain that prefix), then the prefix string may be given to in.pre=. Second, a sequence of column numbers may be given to in.var= with the #:# (if the columns are contiguous) or as a vector (if the columns are not contiguous). Third, a vector of column names may be given to in.var=. Note that one, but not both, of in.var= or in.pre= must be specified by the user.

Note that the computed total radius-at-capture will only be the actual total radius-at-capture if all growth, including plus-growth in the current season, is recorded in the input data frame.

Value

A data.frame of increments in one-fish-per-line (i.e., "wide") format with the total radius-at-capture appended in a variable named as given in var.name (defaults to "radcap").

Author(s)

```
Derek H. Ogle, <derek@derekogle.com>
```

See Also

See gConvert for related functionality.

Examples

```
## Get data with radial measurements
data(SMBassWB)
headtail(SMBassWB)

## convert radial measurements to increments
SMBi1 <- gConvert(SMBassWB,in.pre="anu",out.type="inc")
headtail(SMBi1)

## add the radius-at-capture measurement
SMBi1a <- addRadCap(SMBi1,in.pre="inc",var.name="radcap2")
headtail(SMBi1a)</pre>
```

addZeroCatch 5

addZeroCatch Adds zeroes for catches of species not collected in some sample events.
--

Description

Adds zeroes for catches of species that were not captured in a sampling event but were captured in at least one other sampling event (i.e., adds zeroes to the data frame for capture events where a species was not observed).

Usage

```
addZeroCatch(df, eventvar, specvar, zerovar, na.rm = TRUE)
```

Arguments

df	A data frame that contains the capture summary data as described in the details.
eventvar	A string for the variable that identifies unique capture events.
specvar	A string for the variable that identifies the species captured.
zerovar	A string or vector of strings for the variable(s) that should be set equal to zero. See details.
na.rm	A logicial that indicates if rows of df that are NA should be removed after adding the zeroes. See details.

Details

The data frame in df must contain a column that identifies a unique capture event (given in eventvar), a column with the name for the species captured (given in specvar), and a column that contains the number of that species captured (potentially given to zerovar; see details). All sampling event and species combinations where catch information does not exist is identified and a new data frame that contains a zero for the catch for all of these combinations is created. This new data frame is appended to the original data frame to construct a data frame that contains complete catch information – i.e., including zeroes for species in events where that species was not captured.

The data frame may contain other information related to the catch, such as number of recaptured fish, number of fish released, etc. These additional variables can be includedn in zerovar so that zeroes will be added to these variables as well (e.g., if the catch of the species is zero, then the number of recaptures must also be zero). All variables not given in eventvar, specvar, or zerovar will be assumed to related to eventvar and specvar (e.g., date, gear type, and habitat) and, thus, will be repeated with these variables.

In situations where no fish were captured in some events, the df may contain rows that have a value for eventvar but not for specvar. These rows are important because zeroes need to be added for each observed species for these events. However, in these situations, a <NA> species will appear in the resulting data.frame. It is unlikely that these "missing" species are needed so they will be removed if na.rm=TRUE (default) is used.

6 addZeroCatch

One should test the results of this function by creating a frequency table of the eventvar or specvar. In either case, the table should contain the same value in each cell of the table. See the examples.

Value

A data frame with the same structure as df but with rows of zero observation data appended.

IFAR Chapter

2-Basic Data Manipulations

Note

An error will be returned if either specvar or eventvar are factors with any NA levels. This usually arises if the dataframe was subsetted/filtered prior to using addZeroCatch. See filterD or droplevels for descriptions of how to drop unused levels.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Examples

```
## Example Data #1 (some nets missing some fish, ancillary net data)
df1 <- data.frame(net=c(1,1,1,2,2,3),
                   eff=c(1,1,1,1,1,1),
                   species=c("BKT","LKT","RBT","BKT","LKT","RBT"),
                   catch=c(3,4,5,5,4,3))
df1
xtabs(~net+species,data=df1)
                                                # not all 1s
df1mod1 <- addZeroCatch(df1, "net", "species", zerovar="catch")</pre>
df1mod1
xtabs(~net,data=df1mod1)
                                                # check, should all be 3
xtabs(~net+species,data=df1mod1)
                                               # check, should all be 1
Summarize(catch~species,data=df1mod1)  # correct mean/sd of catches
Summarize(catch~species,data=df1)  # incorrect mean/sd of catche
Summarize(catch~species,data=df1)
                                               # incorrect mean/sd of catches (no zeroes)
# Same as example 1 but with no ancillary data specific to the net number
df2 <- df1[,-2]
df1mod2 <- addZeroCatch(df2, "net", "species", zerovar="catch")</pre>
df1mod2
xtabs(~net+species,data=df1mod2)
                                                # check, should all be 1
## Example Data #3 (All nets have same species ... no zeroes needed)
df3 <- data.frame(net=c(1,1,1,2,2,2,3,3,3),
```

```
eff=c(1,1,1,1,1,1,1,1,1,1),
                  species=c("BKT","LKT","RBT","BKT","LKT","RBT","BKT","LKT","RBT"),
                  catch=c(3,4,5,5,4,3,3,2,7))
df3
xtabs(~net+species,data=df3)
                                              # should all be 1 for this example
df3mod1 <- addZeroCatch(df3,"net","species",zerovar="catch")</pre>
xtabs(~net+species,data=df3mod1)
                                              # check, should still all be 1
## Example Data #4 (another variable that needs zeroes)
df4 <- df1
df4\ensuremath{\text{recaps}} <- c(0,0,0,1,2,1)
df4
xtabs(~net+species,data=df4)
                                              # not all 1s
df4mod1 <- addZeroCatch(df4,"net","species",zerovar=c("catch","recaps"))</pre>
df4mod1
                                             # note zeroes in both variables
xtabs(~net+species,data=df4mod1)
                                              # check, should all be 1
Summarize(catch~species,data=df4)
                                             # observe difference from next
Summarize(catch~species,data=df4mod1)
Summarize(recaps~species,data=df4)
                                              # observe difference from next
Summarize(recaps~species,data=df4mod1)
## Example Data #5 (two "specvar"s)
df5 <- df1
df5$sex <- c("m","m","f","m","f","f")
xtabs(~sex+species+net,data=df5)
                                              # not all 1s
df5mod1 <- addZeroCatch(df5,"net",c("species","sex"),zerovar="catch")</pre>
df5mod1
xtabs(~sex+species+net,data=df5mod1)
                                              # all 1s
str(df5mod1)
## Example Data #6 (three "specvar"s)
df6 <- df5
df6$size <- c("lrg","lrg","lrg","sm","lrg","sm")
df6mod1 <- addZeroCatch(df6,"net",c("species","sex","size"),zerovar="catch")</pre>
df6mod1
```

ageBias

Compute and view possible biases between paired sets of ages.

Description

Constructs age-agreement tables, statistical tests to detect bias, and plots to visualize potential bias in paired age assignments. Ages may be from, for example, two readers of the same structure, one

reader at two times, two stuctures (e.g., scales, spines, otoliths), or a structure and known ages, or one structure and known ages.

Usage

```
ageBias(formula, data, ref.lab = tmp$Enames, nref.lab = tmp$Rname,
 method = stats::p.adjust.methods, sig.level = 0.05, min.n.CI = 3)
## S3 method for class 'ageBias'
summary(object, what = c("table", "symmetry", "Bowker",
 "EvansHoenig", "McNemar", "bias", "diff.bias", "n"), flip.table = FALSE,
 zero.print = "-", digits = 3, cont.corr = c("none", "Yates", "Edwards"),
  ...)
## S3 method for class 'ageBias'
plot(x, what = c("bias", "sunflower", "numbers"),
 difference = FALSE, xlab = x$ref.lab, ylab = x$nref.lab,
  show.n = TRUE, nYpos = 1.03, cex.n = 0.75, lwd = 1,
  show.pts = FALSE, pch.pts = 19, col.pts = grDevices::rgb(0, 0, 0,
  transparency), transparency = 1/10, pch.mean = 95, cex.mean = 1wd,
  col.CI = "black", col.CIsig = "red", lwd.CI = lwd, sfrac = 0,
  show.range = FALSE, col.range = "gray", lwd.range = lwd,
  col.agree = "black", lwd.agree = lwd, lty.agree = 2,
  cex.numbers = 0.9, xlim = NULL, ylim = NULL,
 yaxt = graphics::par("yaxt"), xaxt = graphics::par("xaxt"), ...)
```

Arguments

formula	A formula of the form nrefvar~refvar, where nrefvar and refvar generically represent the variables that contain the paired "nonreference" and "reference" age assignments, respectively. See details.
data	A data frame that minimally contains the paired age assignments given in formula.
ref.lab	A string that contains a label for the reference age assignments.
nref.lab	A string that contains a label for the nonreference age assignments.
method	A string that indicates which method to use for adjusting p-values for multiple comparisons. See ?p.adjust.methods.
sig.level	A numeric value used to determine whether a p-value indicates a significant result. The confidence level used in plot is 100*(1-sig.level).
min.n.CI	A numeric value (default is 3) that indicates the smallest sample size for which a confidence interval should be computed.
what	A string that indicates what type of summary to print or plot to construct. See details.
flip.table	A logical that indicates whether the age-agreement table should be 'flipped' (i.e., rows are reversed so that younger ages are at the bottom of the table). This makes the table more directly comparable to the age-bias plot.
zero.print	A string that indicates what should be printed in place of the zeroes on an age-agreement table. The default is to print a single dash.

digits	A numeric value that indicates the minimum number of digits to print when showing what="bias" or what="diff.bias" in summary.
cont.corr	A string that indicates the continuity correction method to be used with (only) McNemar test. If "none" (default) then no continuity correction is used, if "Yates" then 0.5 is used, and if "Edwards" then 1 is used.
x,object	An object of class ageBias, usually a result from ageBias.
difference	A logical that indicates whether or not the difference between the two age assignments should be used. See details.
xlab,ylab	A string that contains a label for the x-axis (reference) or y-axis (non-reference) age assignments, respectively.
show.n	A logical that indicates whether the sample sizes for each level of the x-axis variable is shown (=TRUE, default) or not (=FALSE).
nYpos	A numeric value that indicates the relative Y position of the sample size values when show.n=TRUE. For example, if nYpos=1.03 then the sample size values will be centered at 3 percent above the top end of the y-axis.
cex.n	A character expansion value for the size of the sample size values.
lwd	A single numeric value that can be used to control the separate 'lwd' argument (e.g., lwd.CI, lwd.range).
show.pts	A logical that indicates whether to show the raw data points.
pch.pts	A numeric value that indicates the plotting character to be used when plotting the raw data points.
col.pts	A string or numeric value that indicates the color to be used for plotting the raw data points. The default is to use black with the transparency found in transparency.
transparency	A numeric value (between 0 and 1) that indicates the level of transparency to use for plotting the raw data points. If expressed as 1/x, then x points plotted on top of each other will represent the color in col.pts.
pch.mean	A numeric value that indicates the plotting character to be used for the mean values (i.e., center of confidence interval bars).
cex.mean	A character expansion value for the size of the mean symbol in pch. mean.
col.CI	A string or numeric value that indicates the color to be used for confidence interval bars that are considered non-significant.
col.CIsig	A string or numeric value that indicates the color to be used for confidence interval bars that are considered significant.
lwd.CI	A numeric value that indicates the line width for the confidence interval bars.
sfrac	A numeric value that controls the size of the ends of the confidence interval bars. See sfrac in plotCI of plotrix .
show.range	A logical that indicates whether to show vertical bars that represent the range of the data points.
col.range	A string or numeric value that indicates the color to be used for the interval representing the range of the data.
lwd.range	A numeric value that indicates the line width for the interval representing the range of the data.

col.agree	A string or numeric value that indicates the color for the 1:1 or zero (if difference) reference line.
lwd.agree	A numeric value that indicates the line width for the 1:1 or zero (if difference) reference line.
lty.agree	A numeric value that indicates the line type for the 1:1 or zero (if difference) reference line.
cex.numbers	A character expansion value for the size of the numbers plotted when what="numbers" is used.
xlim,ylim	A numeric vector of length 2 that contains the limits of the x-axis (reference ages) or y-axis (non-reference ages), respectively.
xaxt,yaxt	A string which specifies the x- and y-axis types. Specifying "n" suppresses plotting of the axis. See ?par.
	Additional arguments for methods.

Details

Generally, one of the two age assessments will be identified as the "reference" set. In some cases this may be the true ages, the ages from the more experience reader, the ages from the first reading, or the ages from the structure gnerally thought to provide the most accurate results. In other cases, such as comparing two novice readers, the choice may be arbitrary. The reference ages will form the columns of the age-agreement table and will be the "constant" age used in the t-tests and age-bias plots (i.e., the x-axis). See further details below.

The age-agreement table is constructed with what="table" in summary. The agreement table can be "flipped", i.e., the rows in descending rather than ascending order, with flip.table=TRUE. By default, the tables are shown with zeroes replaced by dashes. This behavior can be changed with zero.print.

Three statistical tests of symmetry for the age-agreement table can be computed with what= in summary. The "unpooled" or Bowker test as described in Hoenig et al. (1995) is constructed with what="Bowker", the "semi-pooled" or Evans-Hoenig test as described in Evans and Hoenig (1998) is constructed with what="EvansHoenig", and the "pooled" or McNemar test as described in Evans and Hoenig (1998) is constructed with what="McNemar". All three tests are run simultaneously with what="symmetry".

An age-bias plot, as defined by Campana et al. (1995), is constructed with what="bias" (the default) in plot. The reference variable from the ageBias call is plotted on the x-axis. Plotted confidence intervals are computed for the mean of the non-reference ages at each age of the reference ages. The level of confidence is controlled by sig.level= given in the original ageBias call (i.e., confidence level is 100*(1-sig.level). Confidence intervals are only shown if the sample size is greater than the value in min.n.CI=. Confidence intervals plotted in red do not contain the reference age (see discussion of t-tests below). Vertical lines that connect the minimum to the maximum observed value of the y-axis variable at each age of the x-axis variable are plotted in grey if show.range=TRUE. Individual points are plotted if show.pts=TRUE The 1:1 (45 degree) agreement line is shown for comparative purposes. The sample sizes at each age of the x-axis variable are shown if show.n=TRUE (the default). The position of the sample sizes is controlled with nYpos=.

An age-bias plot, as defined by Muir et al. (2008), is constructed as defined above but by also including difference=TRUE in Rplot so that the y-axis is the difference in the paired reference and non-reference ages from the ageBias call (specifically, non-reference-reference).

The frequency of observations at each unique (x,y) coordinate are shown by using what="numbers" in plot.

A "sunflower plot", which contains a symbol for each unique (x,y) coordinate with as many "petals" as observations at that point, is constructed with what="sunflower" in plot. A sunflower plot with differences between the two structures can be constructed by also including difference=TRUE.

Individual t-tests to determine if the mean age of the non-reference set at a particular age of the reference set is equal to the reference age (e.g., is the mean age of the non-reference at age-3 of the reference set statistically equal to 3?) are shown with what="bias" in summary. The results provide a column that indicates whether the difference is significant or not as determined by adjusted p-values from the t-tests and using the signficance level provided in sig.level (defaults to 0.05). Similar results for the difference in ages (e.g., is the mean row variable age minus column variable age at column variable age-3 equal to 0?) are constructed with what="diff.bias" in summary.

The sample size present in the age-agreement table is found with what="n".

Value

ageBias returns a list with the following items:

- data A data.frame with the original paired age assignments and the difference between those assignements.
- agree The age-agreement table.
- bias A data.frame that contains the bias statistics.
- bias.diff A data.frame that contains the bias statistics for the differences.
- ref.lab A string that contains an optional label for the age assignments in the columns (reference) of the age-agreement table.
- nref.lab A string that contains an optional label for the age assignments in the rows (non-reference) of the age-agreement table.

The summary returns the result if what= contains one item, otherwise it returns nothing. Nothing is returned by plot, but see details for a description of the plot that is produced.

Testing

Tested all symmetry test results against results in Evans and Hoenig (2008), the McNemar and Evans-Hoenig results against results from compare2 in **fishmethods**, and all results using the AlewifeLH data set from **FSAdata** against results from http://www.nefsc.noaa.gov/fbp/age-prec/.

IFAR Chapter

4-Age Comparisons.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Campana, S.E., M.C. Annand, and J.I. McMillan. 1995. Graphical and statistical methods for determining the consistency of age determinations. Transactions of the American Fisheries Society 124:131-138.

Evans, G.T. and J.M. Hoenig. 1998. Testing and viewing symmetry in contingency tables, with application to readers of fish ages. Biometrics 54:620-629.

Hoenig, J.M., M.J. Morgan, and C.A. Brown. 1995. Analysing differences between two age determination methods by tests of symmetry. Canadian Journal of Fisheries And Aquatic Systems 52:364-368.

McBride, R.S. 2015. Diagnosis of paired age agreement: A simulation approach of accuracy and precision effects. ICES Journal of Marine Science, XX:XXX-XXX.

Muir, A.M., M.P. Ebener, J.X. He, and J.E. Johnson. 2008. A comparison of the scale and otolith methods of age estimation for lake whitefish in Lake Huron. North American Journal of Fisheries Management 28:625-635.

See Also

See agePrecision for measures of precision between pairs of age assignments. See compare2 in **fishmethods** for similar functionality.

Examples

```
data(WhitefishLC)
ab1 <- ageBias(scaleC~otolithC,data=WhitefishLC,ref.lab="Otolith Age",nref.lab="Scale Age")
summary(ab1)
summary(ab1,what="symmetry")
summary(ab1,what="Bowker")
summary(ab1,what="EvansHoenig")
summary(ab1, what="McNemar")
summary(ab1, what="McNemar", cont.corr="Yates")
summary(ab1,what="bias")
summary(ab1,what="diff.bias")
summary(ab1,what="n")
summary(ab1,what=c("n","symmetry","table"))
# show the zeroes (rather than dashes)
summary(ab1, what="table", zero.print="0")
# flip the table -- ease of comparison to age-bias plot
summary(ab1,what="table",flip.table=TRUE)
## default plot
plot(ab1)
## demonstrates squaring up the axes
plot(ab1, ylim=c(-1, 23), xlim=c(-1, 23))
## plot with the data points shown
plot(ab1, show.pts=TRUE, transparency=1/8)
## plot with the range shown
plot(ab1, show.range=TRUE)
## plot with no difference in significance bar colors
```

ageKey 13

```
plot(ab1,col.CIsig="black")
## plot of differences (note could use same modifications as shown above)
plot(ab1,difference=TRUE)
## sunflower plot
plot(ab1,what="sunflower")
plot(ab1,what="sunflower",difference=TRUE)
## "Numbers" plot
plot(ab1,what="number",col.agree="gray50")
```

ageKey

Use an age-length key to assign age to individuals in the unaged sample.

Description

Use either the semi- or completely-random methods from Isermann and Knight (2005) to assign ages to individual fish in the unaged sample according to the information in an age-length key supplied by the user.

Usage

```
ageKey(key, formula, data, type = c("SR", "CR"), breaks = NULL,
    seed = NULL)

alkIndivAge(key, formula, data, type = c("SR", "CR"), breaks = NULL,
    seed = NULL)
```

Arguments

key	A numeric matrix that contains the age-length key. The format of this matrix is important. See details.
formula	A formula of the form age~length where age generically represents the variable that will contain the estimated ages once the key is applied (i.e., should currently contain no values) and length generically represents the variable that contains the known length measurements. If only ~length is used, then a new variable called "age" will be created in the resulting data frame.
data	A data frame that minimally contains the length measurements and possibly contains a variable that will receive the age assignments as given in formula.
type	A string that indicates whether to use the semi-random (type="SR", default) or completely-random (type="CR") methods for assigning ages to individual fish. See the IFAR chapter for more details.
breaks	A numeric vector of lower values that define the length intervals. See details.
seed	A single numeric that is given to set.seed to set the random seed. This allows repeatability of results.

14 ageKey

Details

The age-length key in key must have length intervals as rows and ages as columns. The row names of key (i.e., rownames (key)) must contain the minimum values of each length interval (e.g., if an interval is 100-109, then the corresponding row name must be 100). The column names of key (i.e., colnames (key)) must contain the age values (e.g., the columns can NOT be named with "age.1", for example).

The length intervals in the rows of key must contain all of the length intervals present in the unaged sample to which the age-length key is to be applied (i.e., sent in the length portion of the formula). If this constraint is not met, then the function will stop with an error message.

If breaks=NULL, then the length intervals for the unaged sample will be determined with a starting interval at the minimum value of the row names in key and a width of the length intervals as determined by the minimum difference in adjacent row names of key. If length intervals of differing widths were used when constructing key, then those breaks should be supplied to breaks=. Use of breaks= may be useful when "uneven" length interval widths were used because the lengths in the unaged sample are not fully represented in the aged sample. See the examples.

Assigned ages will be stored in the column identified on the left-hand-side of formula (if the formula has both a left- and right-hand-side). If this variable is missing in formula, then the new column will be labeled with age.

Value

The original data.frame in data with assigned ages added to the column supplied in formula or in an additional column labeled as age. See details.

Testing

The type="SR" method worked perfectly on a small example. The type="SR" method provides results that reasonably approximate the results from alkAgeDist and alkMeanVar, which suggests that the age assessments are reasonable.

IFAR Chapter

5-Age-Length Key.

Author(s)

Derek H. Ogle, <derek@derekogle.com>. This is largely an R version of the SAS code provided by Isermann and Knight (2005).

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Isermann, D.A. and C.T. Knight. 2005. A computer program for age-length keys incorporating age assignment to individual fish. North American Journal of Fisheries Management, 25:1153-1160.

ageKey 15

See Also

See alkAgeDist and alkMeanVar for alternative methods to derived age distributions and mean (and SD) values for each age. See alkPlot for methods to visualize age-length keys.

Examples

```
## Get data with length measurements and some assigned ages
data(WR79)
## First Example -- Even breaks for length categories
WR1 <- WR79
                                                        # add length categories (width=5)
WR1$LCat <- lencat(WR1$len,w=5)</pre>
WR1.age <- subset(WR1, !is.na(age))</pre>
                                                        # isolate aged and unaged samples
WR1.len <- subset(WR1, is.na(age))</pre>
head(WR1.len)
                                                        # note no ages in unaged sample
raw <- xtabs(~LCat+age,data=WR1.age)</pre>
                                                        # create age-length key
( WR1.key <- prop.table(raw, margin=1) )</pre>
WR1.len <- alkIndivAge(WR1.key,age~len,data=WR1.len) # apply the age-length key
head(WR1.len)
                                                        # now there are ages
WR1.comb <- rbind(WR1.age, WR1.len)</pre>
                                                        # combine orig age & new ages
                                                        # mean length-at-age
Summarize(len~age,data=WR1.comb,digits=2)
( af <- xtabs(~age,data=WR1.comb) )</pre>
                                                        # age frequency distribution
( ap <- prop.table(af) )</pre>
                                                        # proportional age distribution
## Second Example -- length sample does not have an age variable
WR2.age <- subset(WR2, !is.na(age))</pre>
                                                        # isolate age and unaged samples
WR2.len <- subset(WR2, is.na(age))</pre>
WR2.len \leftarrow WR2.len[,-3]
                                                      # remove age variable (for demo only)
WR2.age$LCat <- lencat(WR2.age$len,w=5)</pre>
                                                      # add length categories to aged sample
raw <- xtabs(~LCat+age,data=WR2.age)</pre>
                                                        # create age-length key
( WR2.key <- prop.table(raw, margin=1) )</pre>
WR2.len <- alkIndivAge(WR2.key,~len,data=WR2.len)</pre>
                                                        # apply the age-length key
WR2.len$LCat <- lencat(WR2.len$len,w=5)</pre>
                                                        # add length cat to length sample
head(WR2.len)
                                                        # now there are ages
WR2.comb <- rbind(WR2.age, WR2.len)</pre>
                                                        # combine orig age & new ages
Summarize(len~age,data=WR2.comb,digits=2)
## Third Example -- Uneven breaks for length categories
WR3 <- WR79
brks < c(seq(35,100,5),110,130)
                                                        # set up uneven breaks
WR3$LCat <- lencat(WR3$len,breaks=brks)</pre>
                                                        # add length categories (width=5)
WR3.age <- subset(WR3, !is.na(age))</pre>
                                                        # isolate aged and unaged samples
WR3.len <- subset(WR3, is.na(age))</pre>
head(WR3.len)
                                                        # note no ages in length sample
raw <- xtabs(~LCat+age,data=WR3.age)</pre>
                                                        # create age-length key
( WR3.key <- prop.table(raw, margin=1) )</pre>
WR3.len <- alkIndivAge(WR3.key,age~len,data=WR3.len,breaks=brks) # apply the age-length key
head(WR3.len)
                                                        # now there are ages
WR3.comb <- rbind(WR3.age, WR3.len)</pre>
                                                        # combine orig age & new ages
Summarize(len~age,data=WR3.comb,digits=2)
```

16 ageKeyPlot

ageKeyPlot	Plots to visualize age-length keys.	

Description

Various plots to visualize the proportion of fish of certain ages within length intervals in an agelength key.

Usage

```
ageKeyPlot(key, type = c("barplot", "area", "lines", "splines", "bubble"),
    xlab = "Length", ylab = ifelse(type != "bubble", "Proportion", "Age"),
    xlim = NULL, ylim = NULL, showLegend = FALSE, lbl.cex = 1.25,
    leg.cex = 1, lwd = 2, span = 0.25, pal = paletteChoices(),
    grid = TRUE, col = "gray80", buf = 0.45, add = FALSE, ...)

alkPlot(key, type = c("barplot", "area", "lines", "splines", "bubble"),
    xlab = "Length", ylab = ifelse(type != "bubble", "Proportion", "Age"),
    xlim = NULL, ylim = NULL, showLegend = FALSE, lbl.cex = 1.25,
    leg.cex = 1, lwd = 2, span = 0.25, pal = paletteChoices(),
    grid = TRUE, col = "gray80", buf = 0.45, add = FALSE, ...)
```

Arguments

key	A numeric matrix that contains the age-length key.
type	A string that indicates the type of plot to construct. See details.
xlab,ylab	A string that contains the label for the x- or y-axis.
xlim,ylim	A numeric of length 2 that provide the limits for the x-axis or y-axis.
showLegend	A logical that indicates whether a legend should be displayed (not implemented for type="bubble"). See examples.
lbl.cex	A numeric character expansion value for labels inside the bars when type="barplot" or on the lines when type="lines" or type="splines". Only used if showLegend=FALSE.
leg.cex	A numeric character expansion value for labels on the legend when showLegend=TRUE.
lwd	A numeric that indicates the line width when type="lines" or type="splines".
span	A numeric that indicates the span value to use in loess when type="splines".
pal	A string that indicates the palette to generate colors for the bars, areas, lines, or spline lines. The name of a palette must be one of "rich", "cm", "default", "grey", "gray", "heat", "jet", "rainbow", "topo", or "terrain". See chooseColors.
grid	A logical that indicates whether a grid should be placed under the bubbles when type="bubble" or a character or appropriate vector that identifies a color for the grid. See examples.
col	A string that indicates the color of the bubbles when type="bubble".

ageKeyPlot 17

buf A single numeric that indicates the relative width of the bubbles when type="bubble".

A value of 0.5 means that two full-width bubbles would touch each other either in the x- or y-direction (i.e., this would represent half of the minimum of the physical distance between values one-unit apart on the x- and y-axes). Set this to a value less than 0.5 so that the bubbles will not touch (the default is 0.45).

A logical that indicates whether the data should be added to an already existing

plot. May be useful for visually comparing age-length keys. Only implemented

when type="bubble".

. . . Additional arguments to pass to plot or barplot.

Details

A variety of plots can be used to visualize the proportion of fish of certain ages within length intervals of an age-length key. The types of plots are described below and illustrated in the examples.

- A "stacked" bar chart where vertical bars over length intervals sum to 1 but are segmented by the proportion of each age in that length interval is constructed with type="barplot". The ages will be labeled in the bar segments unless showLegend=TRUE is used.
- A "stacked" area chart similar to the bar chart described above is constructed with type="area".
- A plot with (differently colored) lines that connect the proportions of ages within each length interval is constructed with type="lines".
- A plot with (differently colored) lines, as estimated by loess splines, that connect the proportions of ages within each length interval is constructed with type="splines".
- A "bubble" plot where circles whose size is proportional to the proportion of fish of each age
 in each length interval is constructed with type="bubble". The color of the bubbles can be
 controlled with col= and an underlying grid for ease of seeing the age and length interval
 for each bubble can be controlled with grid=. Bubbles from a second age-length key can be
 overlaid on an already constructed bubble plot by using add=TRUE in a second call to alkPlot.

Note that all plots are "vertically conditional" – i.e., each represents the proportional ages WITHIN each length interval.

Value

None, but a plot is constructed.

IFAR Chapter

5-Age-Length Key.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

18 agePrecision

See Also

See alkIndivAge for using an age-length key to assign ages to individual fish.

Examples

```
## Make an example age-length key -- same as in ageKey()
data(WR79)
WR.age <- subset(WR79, !is.na(age))</pre>
                                           # isolate the age sample
WR.age$LCat <- lencat(WR.age$len,w=5)</pre>
                                           # add length intervals (width=5)
raw <- xtabs(~LCat+age,data=WR.age)</pre>
                                           # create age-length key
( WR.key <- prop.table(raw, margin=1) )</pre>
## Various visualizations of the age-length key
alkPlot(WR.key, "barplot")
alkPlot(WR.key, "barplot", pal="gray")
alkPlot(WR.key,"barplot",showLegend=TRUE)
alkPlot(WR.key, "area")
alkPlot(WR.key, "area", showLegend=TRUE)
alkPlot(WR.key, "area", pal="gray")
alkPlot(WR.key,"lines")
alkPlot(WR.key,"lines",pal="gray")
alkPlot(WR.key, "lines", showLegend=TRUE)
alkPlot(WR.key, "splines")
alkPlot(WR.key, "splines", span=0.2)
alkPlot(WR.key, "splines", pal="gray", showLegend=TRUE)
alkPlot(WR.key,"bubble")
alkPlot(WR.key, "bubble", grid=FALSE)
alkPlot(WR.key, "bubble", grid="blue")
alkPlot(WR.key,"bubble",grid=rgb(0,0,0,0.2),col=rgb(0,0,0,0.5))
```

agePrecision

Compute measures of precision among sets of ages.

Description

Computes overall measures of precision for multiple age assignments made on the same individuals. Ages may be from two or more readers of the same structure, one reader at two or more times, or two or more stuctures (e.g., scales, spines, otoliths). Measures of precision include ACV (Average Coefficient of Variation), APE (Average Percent Error), and various percentage difference values.

Usage

```
agePrecision(formula, data)
## S3 method for class 'agePrec'
summary(object, what = c("precision", "difference",
    "absolute difference", "details"), percent = TRUE, trunc.diff = NULL,
    digits = 4, ...)
```

agePrecision 19

Arguments

formula	A formula of the form ~var1+var2+var3+ or, alternatively, var1~var2+var3+, where the varX generically represent the variables that contain the age assignments. The alternative formula allows for similar code as used in ageBias and can have only one variable on the left-hand side.
data	A data frame that minimally contains the variables in formula.
object	An object of class agePrec, usually from agePrecision.
what	A string (or vector of strings) that indicates what type of summary to print. See details.
percent	A logical that indicates whether the difference table (see details) should be represented as percentages (TRUE; default) or frequency (FALSE) of fish.
trunc.diff	A single integer that identifies the age for which all values that age and greater are combined into one category. See the examples.
digits	A single numeric that indicates the minimum number of digits to print when using summary.
	Additional arguments for methods.

Details

If what="precision" in summary then a summary table that contains the following items will be printed:

- n Number of fish in data.
- R Number of age assessments given in formula.
- ACV The mean coefficient of variation. See the IFAR chapter for calculational details.
- APE The mean average percent error. See the IFAR chapter for calculational details.
- PercAgree The percentage of fish for which all age assignments perfectly agree.

If what="difference" is used in summary then a table that describes either the percentage (if percent=TRUE, default) or frequency of fish by the difference in paired age assignments. This table has one row for each possible pair of age assignments.

If what="absolute difference" is used in summary then a table that describes either the percentage (if percent=TRUE, default) or frequency of fish by the absolute value of the difference in paired age assignments. This table has one row for each possible pair of age assignments. The "1" column, for example, represents age assignments that disagree by one year (in either direction).

If what="detail" is used in summary then a data frame of the original data along with the intermediate caculations of the average age, standard deviation of age, APE, and ACV for each individual will be printed. These details are generally only used to check or to understand calculations.

Value

The main function returns a list with the following items:

 detail A data.frame with all data given in data and intermediate calculations for each fish. See details 20 agePrecision

- rawdiff A frequency table of fish by differences for each pair of ages.
- absdiff A frequency table of fish by absolute differences for each pair of ages.
- APE The mean average percent error.
- ACV The mean coefficient of variation.
- n Number of fish in data.
- R Number of age assessments for each fish given in formula.

The summary returns the result if what= contains one item, otherwise it returns nothing. See details for what is printed.

Testing

Tested all precision results against published results in Herbst and Marsden (2011) for the WhitefishLC data and the results for the AlewifeLH data set from **FSAdata** against results from http://www.nefsc.noaa.gov/fbp/age-prec/.

IFAR Chapter

4-Age Comparisons.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Beamish, R.J. and D.A. Fournier. 1981. A method for comparing the precision of a set of age determinations. Canadian Journal of Fisheries and Aquatic Sciences 38:982-983.

Campana, S.E. 1982. Accuracy, precision and quality control in age determination, including a review of the use and abuse of age validation methods. Journal of Fish Biology 59:197-242.

Campana, S.E., M.C. Annand, and J.I. McMillan. 1995. Graphical and statistical methods for determining the consistency of age determinations. Transactions of the American Fisheries Society 124:131-138.

Chang, W.Y.B. 1982. A statistical method for evaluating the reproducibility of age determination. Canadian Journal of Fisheries and Aquatic Sciences 39:1208-1210.

McBride, R.S. 2015. Diagnosis of paired age agreement: A simulation approach of accuracy and precision effects. ICES Journal of Marine Science, XX:XXX-XXX.

See Also

See ageBias for computation of the full age agreement table, along with tests and plots of age bias.

alkAgeDist 21

Examples

```
## Example with just two age assignments
data(WhitefishLC)
ap1 <- agePrecision(~otolithC+scaleC,data=WhitefishLC)</pre>
summary(ap1)
summary(ap1,what="precision")
summary(ap1,what="difference")
summary(ap1,what="difference",percent=FALSE)
summary(ap1,what="absolute")
summary(ap1, what="absolute", percent=FALSE)
summary(ap1,what="absolute",trunc.diff=4)
summary(ap1, what="absolute", percent=FALSE)
summary(ap1,what=c("precision","difference"))
barplot(ap1$rawdiff,ylab="Frequency",xlab="Otolith - Scale Age")
summary(ap1, what="detail")
## Example with three age assignments
ap2 <- agePrecision(~otolithC+finrayC+scaleC,data=WhitefishLC)</pre>
summary(ap2)
summary(ap2,what="precision")
summary(ap2, what="difference")
summary(ap2,what="difference",percent=FALSE)
summary(ap2,what="absolute")
summary(ap2,what="absolute",percent=FALSE)
summary(ap2,what="absolute",trunc.diff=4)
summary(ap2,what="absolute",percent=FALSE,trunc.diff=4)
summary(ap2,what="detail")
```

alkAgeDist

Proportions-at-age from an Age-Length Key

Description

Computes the proportions-at-age (with standard errors) in a larger sample based on an age-length-key created from a subsample of ages through a two-stage random sampling design. Follows the methods in Quinn and Deriso (1999).

Usage

```
alkAgeDist(key, lenA.n, len.n)
```

Arguments

key	A numeric matrix that contains the age-length key. See details.
lenA.n	A numeric vector of sample sizes for each length interval in the aged sample.
len.n	A numeric vector of sample sizes for each length interval in the <i>complete sample</i> (i.e., all fish regardless of whether they were aged or not).

22 alkAgeDist

Details

The age-length key in key must have length intervals as rows and ages as columns. The row names of key (i.e., rownames (key)) must contain the mininum values of each length interval (e.g., if an interval is 100-109 then the corresponding row name must be 100). The column names of key (i.e., colnames (key)) must contain the age values (e.g., the columns can NOT be named with "age.1", for example).

The length intervals in the rows of key must contain all of the length intervals present in the larger sample. Thus, the length of len.n must, at least, equal the number of rows in key. If this constraint is not met, then the function will stop with an error message.

The values in lenA.n are equal to what the row sums of key would have been before key was converted to a row proportions table. Thus, the length of lenA.n must also be equal to the number of rows in key. If this constraint is not met, then the function will stop with an error message.

Value

A data.frame with as many rows as ages (columns) present in key and the following three variables:

- · age The ages.
- prop The proportion of fish at each age.
- se The SE for the proportion of fish at each age.

Testing

The results from this function perfectly match the results in Table 8.4 (left) of Quinn and Deriso (1999) using SnapperHG2 from **FSAdata**. The results also perfectly match the results from using alkprop in **fishmethods**.

IFAR Chapter

5-Age-Length Key.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Lai, H.-L. 1987. Optimum allocation for estimating age composition using age-length key. Fishery Bulletin, 85:179-185.

Lai, H.-L. 1993. Optimum sampling design for using the age-length key to estimate age composition of a fish population. Fishery Bulletin, 92:382-388.

Quinn, T. J. and R. B. Deriso. 1999. Quantitative Fish Dynamics. Oxford University Press, New York, New York. 542 pages.

alkMeanVar 23

See Also

See alkIndivAge and related functions for a completely different methodology. See alkprop from **fishmethods** for the exact same methodology but with a different format for the inputs.

Examples

```
## Get data with length measurements and some assigned ages
data(WR79)
## Example -- Even breaks for length categories
WR1 <- WR79
# add length intervals (width=5)
WR1$LCat <- lencat(WR1$len,w=5)</pre>
# get number of fish in each length interval in the entire sample
len.n <- xtabs(~LCat,data=WR1)</pre>
# isolate aged sample and get number in each length interval
WR1.age <- subset(WR1, !is.na(age))</pre>
lenA.n <- xtabs(~LCat,data=WR1.age)</pre>
# create age-length key
raw <- xtabs(~LCat+age,data=WR1.age)</pre>
( WR1.key <- prop.table(raw, margin=1) )</pre>
# use age-length key to estimate age distribution of all fish
alkAgeDist(WR1.key,lenA.n,len.n)
```

alkMeanVar

Mean Values-at-age from an Age-Length Key

Description

Computes the mean value-at-age in a larger sample based on an age-length-key created from a subsample of ages through a two-stage random sampling design. The mean values could be mean length-, weight-, or fecundity-at-age, for example. The methods of Bettoli and Miranda (2001) or Quinn and Deriso (1999) are used. A standard deviation is computed for the Bettoli and Miranda (2001) method and standard error for the Quinn and Deriso (1999) method. See the testing section notes.

Usage

```
alkMeanVar(key, formula, data, len.n, method = c("BettoliMiranda",
   "QuinnDeriso"))
```

Arguments

key

A numeric matrix that contains the age-length key. See details.

formula

A formula of the form var~lencat+age where var generically represents the variable to be summarized (e.g., length, weight, fecundity), lencat generically represents the variable that contains the length intervals, and age generically represents the variable that contains the assigned ages.

24 alkMeanVar

data	A data frame that minimally contains the length intervals, assessed ages, and the variable to be summarized (i.e., this should be the aged sample) as given in formula.
len.n	A vector of sample sizes for each length interval in the <i>complete sample</i> (i.e., all fish regardles of whether they were aged or not).
method	A string that indicates which method of calculation should be used. See details.

Details

The age-length key key must have length intervals as rows and ages as columns. The row names of key (i.e., rownames (key)) must contain the minimum values of each length interval (e.g., if an interval is 100-109, then the corresponding row name must be 100). The column names of key (i.e., colnames (key)) must contain the age values (e.g., the columns can NOT be named with "age.1", for example).

The length intervals in the rows of key must contain all of the length intervals present in the larger sample. Thus, the length of len.n must, at least, equal the number of rows in key. If this constraint is not met, then the function will stop with an error message.

Note that the function will stop with an error if the formula in formula does not meet the specific criteria outlined in the parameter list above.

Value

A data.frame with as many rows as ages (columns) present in key and the following three variables:

- age The ages.
- mean The mean value at each age.
- sd,se The SD if method="BettoliMiranda" or SE of the mean if method="QuinnDeriso" for the value at each age.

Testing

The results of these functions have not yet been rigorously tested. The Bettoli and Miranda (2001) results appear, at least, approximately correct when compared to the results from alkIndivAge. The Quinn and Deriso (1999) results appear at least approximately correct for the mean values, but do not appear to be correct for the SE values. Thus, a note is returned with the Quinn and Deriso (1999) results that the SE should not be trusted.

IFAR Chapter

5-Age-Length Key.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

bcFuns 25

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Bettoli, P. W. and Miranda, L. E. 2001. A cautionary note about estimating mean length at age with subsampled data. North American Journal of Fisheries Management, 21:425-428.

Quinn, T. J. and R. B. Deriso. 1999. Quantitative Fish Dynamics. Oxford University Press, New York, New York. 542 pages

See Also

See alkIndivAge and related functions for a completely different methodology. See alkAgeDist for a related method of determining the proportion of fish at each age. See the **ALKr** package.

Examples

```
## Get data with length measurements and some assigned ages
data(WR79)
## Example -- Even breaks for length categories
WR1 <- WR79
# add length intervals (width=5)
WR1$LCat <- lencat(WR1$len,w=5)</pre>
# get number of fish in each length interval in the entire sample
len.n <- xtabs(~LCat,data=WR1)</pre>
# isolate aged sample
WR1.age <- subset(WR1, !is.na(age))</pre>
# create age-length key
raw <- xtabs(~LCat+age,data=WR1.age)</pre>
( WR1.key <- prop.table(raw, margin=1) )</pre>
## use age-length key to estimate mean length-at-age of all fish
# Bettoli-Miranda method
alkMeanVar(WR1.key,len~LCat+age,WR1.age,len.n)
# Quinn-Deriso method
alkMeanVar(WR1.key,len~LCat+age,WR1.age,len.n,method="QuinnDeriso")
```

bcFuns

Creates a function for a specific back-calculation model.

Description

Creates a function for a specific back-calculation model based on definitions in Vigloila and Meekan (2009).

Usage

```
bcFuns(BCM, msg = FALSE)
```

26 bcFuns

Arguments

BCM	A single numeric between 1 and 22 or a string that indicates which back-calculation model to use (based on numbers and names in Vigliola and Meekan (2009)).
msg	A logical that indicates whether a message about the model and parameter definitions should be output.

Details

The following back-calculation models, based on definitions with abbreviations and model numbers from Vigloila and Meekan (2009), are supported.

Abbreviation	Number	Model
DALE	1	Dahl-Lea
FRALE	2	Fraser-Lee
BI, LBI	3	(Linear) Biological Intercept
BPH, LBPH	4	(Linear) Body Proportional Hypothesis
TVG	5	Time-Varying Growth
SPH, LSPH	6	(Linear) Scale Proportional Hypothesis
AE, AESPH	7	(Age Effect) Scale Proportional Hypothesis
AEBPH	8	(Age Effect) Body Proportional Hypothesis
MONA	9	Monastyrsky
MONA-BPH	10	Monastyrsky Body Proportional Hypothesis
MONA-SPH	11	Monastyrsky Scale Proportional Hypothesis
WAKU	12	Watanabe and Kuroki
FRY	13	Fry
MF, ABI	14	Modified Fry, Allometric Biological Intercept
FRY-BPH, ABPH	15	Fry, Allometric Body Proportional Hypothesis
FRY-SPH, ASPH	16	Fry, Allometric Scale Proportional Hypothesis
QBPH	17	Quadratic Body Proportional Hypothesis
QSPH	18	Quadratic Scale Proportional Hypothesis
PBPH	19	Polynomial Body Proportional Hypothesis
PSPH	20	Polynomial Scale Proportional Hypothesis
EBPH	21	Exponential Body Proportional Hypothesis
ESPH	22	Exponential Scale Proportional Hypothesis

Value

A function that can be used to predict length at previous age (Li) given length-at-capture (Lc), hard-part radius-at-age i (Ri), and hard-part radius-at-capture (Rc). In addition, some functions/models may require the previous age (agei) and the age-at-capture (agec), certain parameters related to the biological intercept (R0p & L0p), or certain parameters estimated from various regression models (a,b,c,A,B,C). See source for more information.

fishR vignette

https://sites.google.com/site/fishrfiles/gnrl/Backcalculation.pdf,https://sites.google.com/site/fishrfiles/gnrl/BackcalculationExtra.pdf

binCI 27

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Vigliola, L. and M.G. Meekan. 2009. The back-calculation of fish growth from otoliths. pp. 174-211. in B.S. Green et al. (editors). Tropical Fish Otoliths: Information for Assessment, Management and Ecology. Review: Methods and Technologies in Fish Biology and Fisheries 11. Springer.

Examples

binCI

Confidence intervals for binomial probability of success.

Description

Uses one of three methods to compute a confidence interval for the probability of success (p) in a binomial distribution.

Usage

```
binCI(x, n, conf.level = 0.95, type = c("wilson", "exact", "asymptotic",
    "all"))
```

Arguments

x A single or vector of numbers that contains the number of observed successes.

n A single or vector of numbers that contains the sample size.

conf.level A single number that indicates the level of confidence (default is 0.95).

type A string that identifies the type of method to use for the calculations. See details.

Details

This function will compute confidence interval for three possible methods chosen with the type argument.

28 binCI

type="wilson" Wilson's (Journal of the American Statistical Association, 1927) confidence interval for a proportion.
type="exact" Computes the Clopper/Pearson exact CI for a binomial success probability.
type="asymptotic" This uses the normal distribution approximation.

Note that Agresti and Coull (2000) suggest that the Wilson interval is the preferred method and is, thus, the default type.

Value

A #x2 matrix that contains the lower and upper confidence interval bounds as columns.

Note

This is primarily a wrapper function for binconf in the **Hmisc** package (this implementation uses arguments, specificially conf.level, that more closely match other functions).

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Agresti, A. and B.A. Coull. 1998. Approximate is better than "exact" for interval estimation of binomial proportions. American Statistician, 52:119-126.

See Also

See binom. test, binconf in **Hmisc**, binom. conf. int in **epitools**, and functions in **binom**.

Examples

```
binCI(7,20,type="wilson")
binCI(7,20,type="exact")
binCI(7,20,type="asymptotic")

## Demonstrates using all types at once
binCI(7,20,type="all")

## Demonstrates use with multiple inputs
binCI(c(7,10),c(20,30))
```

BluegillJL 29

BluegillJL

Capture histories (2 samples) of Bluegill from Jewett Lake.

Description

Each line consists of the capture history over two samples of Bluegill (*Lepomis macrochirus*) in Jewett Lake (MI). This file contains the capture histories for only Bluegill larger that 6-in.

Format

A data frame with 277 observations on the following 2 variables.

first a numeric vector of indicator variables for the first sample (1=captured).

second a numeric vector of indicator variables for the second sample (1=captured).

Topic(s)

- · Population size
- Abundance
- Mark-recapture
- · Petersen method
- · Capture history

Source

From example 8.1 in Schneider, J.C. 1998. Lake fish population estimates by mark-and-recapture methods. Chapter 8 in Schneider, J.C. (ed.) 2000. Manual of fisheries survey methods II: with periodic updates. Michigan Department of Natural Resources, Fisheries Special Report 25, Ann Arbor.

Examples

```
data(BluegillJL)
str(BluegillJL)
head(BluegillJL)
```

30 bootstrap

bootstrap

Associated S3 methods for bootCase from car.

Description

Provides S3 methods to construct non-parametric bootstrap confidence intervals, hypothesis tests, and plots of the parameter estimates for bootCase objects from the **car** package.

Usage

```
## S3 method for class 'bootCase'
confint(object, parm = NULL, level = conf.level,
    conf.level = 0.95, plot = FALSE, err.col = "black", err.lwd = 2,
    rows = NULL, cols = NULL, ...)

## S3 method for class 'bootCase'
predict(object, FUN, MARGIN = 1, conf.level = 0.95,
    digits = NULL, ...)

## S3 method for class 'bootCase'
htest(object, parm = NULL, bo = 0, alt = c("two.sided",
    "less", "greater"), plot = FALSE, ...)

## S3 method for class 'bootCase'
hist(x, same.ylim = TRUE, ymax = NULL,
    rows = round(sqrt(ncol(x))), cols = ceiling(sqrt(ncol(x))), ...)

## S3 method for class 'bootCase'
plot(x, ...)
```

Arguments

object,x	A bootCase object.
parm	A number or string that indicates which column of object contains the parameter estimates to use for the confidence interval or hypothesis test.
level	Same as conf.level.
conf.level	A level of confidence as a proportion.
plot	A logical that indicates whether a plot should be constructed. If confint then a histogram of the parm parameters from the bootstrap samples with error bars that illustrate the bootstrapped confidence intervals will be constructed. If codehtest then a histogram of the parm parameters with a vertical line illustrating the bo value will be constructed.
err.col	A single numeric or character that identifies the color for the error bars on the plot.
err.lwd	A single numeric that identifies the line width for the error bars on the plot.

bootstrap 31

rows A single numeric that contains the number of rows to use on the graphic. A single numeric that contains the number of columns to use on the graphic. cols The function to be applied for the prediction. See the examples. FUN A single numeric that indicates the margin over which FUN is applied. MARGIN=1 MARGIN will apply to each row and is the default. A single numeric that indicates the number of digits for the result. digits bo The null hypothesized parameter value. alt A string that indicates the "direction" of the alternative hypothesis. See details. A logical that indicates whether the same limits for the y-axis should be used on same.ylim each histogram. Defaults to TRUE. Ignored if ylmts is non-null. A single value that sets the maximum y-axis limit for each histogram or a vector ymax of length equal to the number of groups that sets the maximum y-axis limit for each histogram separately.

... Additional items to send to functions.

A named color for the histogram bars.

Details

col

confint finds the two quantiles that have the (1-conf.level)/2 proportion of bootstrapped parameter estimates below and above. This is an approximate 100conf.level% confidence interval.

predict applies a user-supplied function to each row of object and then finds the median and the two quantiles that have the proportion (1-conf.level)/2 of the bootstrapped predictions below and above. The median is returned as the predicted value and the quantiles are returned as an approximate 100conf.level% confidence interval for that prediction.

In htest the "direction" of the alternative hypothesis is identified by a string in the alt= argument. The strings may be "less" for a "less than" alternative, "greater" for a "greater than" alternative, or "two.sided" for a "not equals" alternative (the DEFAULT). In the one-tailed alternatives the p-value is the proportion of bootstrapped parameter estimates in object\$coefboot that are extreme of the null hypothesized parameter value in bo. In the two-tailed alternative the p-value is twice the smallest of the proportion of bootstrapped parameter estimates above or below the null hypothesized parameter value in bo.

Value

If object is a matrix, then confint returns a matrix with as many rows as columns (i.e., parameter estimates) in object and two columns of the quantiles that correspond to the approximate confidence interval. If object is a vector, then confint returns a vector with the two quantiles that correspond to the approximate confidence interval.

htest returns a two-column matrix with the first column containing the hypothesized value sent to this function and the second column containing the corresponding p-value.

hist constructs histograms of the bootstrapped parameter estimates.

plot constructs scatterplots of all pairs of bootstrapped parameter estimates.

predict returns a matrix with one row and three columns, with the first column holding the predicted value (i.e., the median prediction) and the last two columns holding the approximate confidence interval.

32 BrookTroutTH

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

S. Weisberg (2005). Applied Linear Regression, third edition. New York: Wiley, Chapters 4 and 11

See Also

bootCase in car.

Examples

```
data(Ecoli)
fnx <- function(days,B1,B2,B3) {</pre>
  if (length(B1) > 1) {
    B2 <- B1[2]
    B3 <- B1[3]
    B1 <- B1[1]
  B1/(1+exp(B2+B3*days))
}
nl1 \leftarrow nls(cells \sim fnx(days, B1, B2, B3), data = Ecoli, start = list(B1 = 6, B2 = 7.2, B3 = -1.45))
if (require(car)) {  # for bootCase()
  nl1.boot <- car::bootCase(nl1,B=99) # B=99 too small to be useful</pre>
  confint(nl1.boot, "B1")
  confint(nl1.boot, c(2,3))
  confint(nl1.boot,conf.level=0.90)
  predict(nl1.boot,fnx,days=3)
  htest(nl1.boot,1,bo=6,alt="less")
  hist(nl1.boot)
  plot(nl1.boot)
  cor(nl1.boot)
}
```

BrookTroutTH

Catch-at-age for Tobin Harbor, Isle Royale Brook Trout.

Description

Catch-at-age in fyke nets from 1996-1998 for "Coaster" Brook Trout (*Salvelinus fontinalis*) in Tobin Harbor, Isle Royale, Lake Superior.

Format

A data frame with 7 observations on the following 2 variables.

age A numeric vector of assigned ages.

catch A numeric vector of number of Brook Trout caught.

Topic(s)

- Mortality
- · Catch curve
- · Chapman-Robson

Source

Quinlan, H.R. 1999. Biological Characteristics of Coaster Brook Trout at Isle Royale National Park, Michigan, 1996-98. U.S. Fish and Wildlife Service Ashland Fishery Resources Office report. November 1999. Was originally available at www.fws.gov/midwest/ashland/brook/biochar/biolchar.html.

Examples

```
data(BrookTroutTH)
str(BrookTroutTH)
head(BrookTroutTH)
plot(log(catch)~age,data=BrookTroutTH)
```

capHistConvert

Convert between capture history data.frame formats.

Description

Use to convert between simple versions of several capture history data.frame formats – "individual", "frequency", "event", "MARK", and "RMark". The primary use is to convert to the "individual" format for use in capHistSum.

Usage

```
capHistConvert(df, cols2use = NULL, cols2ignore = NULL,
  in.type = c("frequency", "event", "individual", "MARK", "marked", "RMark"),
  out.type = c("individual", "event", "frequency", "MARK", "marked", "RMark"),
  id = NULL, event.ord = NULL, freq = NULL, var.lbls = NULL,
  var.lbls.pre = "event", include.id = ifelse(is.null(id), FALSE, TRUE))
```

Arguments

df	A data frame that contains the capture histories and, perhaps, a unique fish identifier or frequency variable. See details.
cols2use	A string or numeric vector that indicates columns in df to use. Negative numeric values will not use those columns. Cannot use both cols2use and col2ignore.
cols2ignore	A string or numeric vector that indicates columns in df to ignore. Typical columns to ignore are those that are not either in id= or freq= or part of the capture history data. Cannot use both cols2use and col2ignore.
in.type	A single string that indicates the type of capture history format to convert FROM .

out.type	A single string that indicates the type of capture history format to convert TO .
id	A string or numeric that indicates the column in df that contains the unique identifier for an individual fish. This argument is only used if in.type="event", in.type="individual", or, possibly, in.type="RMark".
event.ord	A string that contains a vector of ordered levels to be used when in.type="event". The default is to order alphabetically which may not be desirable if, for example, the events are labelled as 'first', 'second', 'third', and 'fourth'. In this case, use event.ord=c("first", "second", "third", "fourth").
freq	A string or numeric that indicates the column in df that contains the frequency of individual fish corresponding to a capture history. This argument is only used if in.type="MARK", in.type="frequency", or, possibly, in.type="RMark".
var.lbls	A string vector of labels for the columns that contain the returned individual or frequency capture histories. If var.lbls=NULL or the length is different then the number of events then default labels using var.lbls.pre will be used. This argument is only used if out.type="frequency" or out.type="individual".
var.lbls.pre	A single string used as a prefix for the labels of the columns that contain the returned individual or frequency capture histories. This prefix will be appended with a number corresponding to the sample event. This argument is only used if out.type="frequency" or out.type="individual" and will be ignored if a proper vector is given in var.lbls.
include.id	A logical that indicates whether a unique fish identifier variable/column should be included in the output data.frame. This argument is only used if out.type="individual" or out.type="RMark".

Details

capHistSum requires capture histories to be recorded in the "individual" format. In this format, the data frame contains (at least) as many columns as sample events and as many rows as individually tagged fish. Optionally, the data.frame may also contain a column with unique fish identifiers (e.g., tag numbers). Each cell in the capture history portion of the data.frame contains a '0' if the fish of that row was NOT seen in the event of that column and a '1' if the fish of that row WAS seen in the event of that column. For example, suppose that five fish were marked on four sampling events; fish '17' was captured on the first two events; fish '18' was captured on the first and third events; fish '19' was captured on only the third event; fish '20' was captured on only the fourth event; and fish '21' was captured on the first and second events. The "individual" capture history date.frame for these data looks like:

fish	event1	event2	event3	event4
17	1	1	0	0
18	1	0	1	0
19	0	0	1	0
20	0	0	0	1
21	1	1	0	0

The "frequency" format data.frame (this format is used in Rcapture) has unique capture histories in separate columns, as in the "individual" format, but also includes a column with the frequency of

individuals that had the capture history of that row. It will not contain a fish identifier variable. The same data from above looks like:

event1	event2	event3	event4	freq
1	1	0	0	2
1	0	1	0	1
0	0	1	0	1
0	0	0	1	1

The "event" format data.frame has a column with the unique fish identifier and a column with the event in which the fish of that row was observed. The same data from above looks like:

fish	even
17	1
18	1
21	1
17	2
21	2
18	3
19	3
20	4

MARK is the "gold-standard" software for analyzing complex capture history information. In the "MARK" format the 0s and 1s of the capture histories are combined together as a string without any spaces. Thus, the "MARK" format has the capture history strings in one column with an additional column that contains the frequency of individuals that exhibited the capture history of that row. The final column ends with a semi-colon. The same data from above looks like:

ch	freq
0001	1;
0010	1;
1010	1;
1100	2;

The RMark and **marked** are packages used to replace some of the functionality of MARK or to interact with MARK. The "RMark" or "marked" format requires the capture histories as one string (must be a character string and called 'ch'), as in the "MARK" format, but without the semicolon. The data.frame may be augmented with an indentifier for individual fish OR with a frequency variable. If augmented with a unique fish identification variable then the same data from above looks like:

fish	ch
17	1100
18	1010
19	0010

However, if augmented with a frequency variable then the same data from above looks like:

ch	freq
0001	1
0010	1
1010	1
1100	2

Each of the formats can be used to convert from (i.e., in in.type=) or to convert to (i.e., in out.type=) with the exception that only the individual fish identifier version can be converted to when out.type="RMark".

Value

A data frame of the proper type given in out. type is returned. See details.

Warning

capHistConvert may give unwanted results if the data are in.type="event" but there are unused levels for the variable, as would result if the data.frame had been subsetted on the event variable. The unwanted results can be corrected by using droplevels before capHistConvert. See the last example for an example.

IFAR Chapter

9-Abundance from Capture-Recapture Data.

Note

The formats as used here are simple in the sense that one is only allowed to have the individual fish identifier or the frequency variable in addition to the capture history information. More complex analyses may use a number of covariates. For these more complex analyses, one should work directly with the Rcapture, RMark, or **marked** packages.

This function also assumes that all unmarked captured fish are marked and returned to the population (i.e., no losses at the time of marking are allowed).

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

capHistConvert 37

See Also

See capHistSum to summarize "individual" capture histories into a format useable in mrClosed and mrOpen. Also see Rcapture, RMark, or **marked** packages for handling more complex analyses.

```
## A small example of 'event' format
(ex1 <- data.frame(fish=c(17,18,21,17,21,18,19,20),yr=c(1987,1987,1987,1988,1988,1989,1989,1990)))
# convert to 'individual' format
( ex1.E2I <- capHistConvert(ex1,id="fish",in.type="event") )</pre>
# convert to 'frequency' format
( ex1.E2F <- capHistConvert(ex1,id="fish",in.type="event",out.type="frequency") )</pre>
# convert to 'MARK' format
( ex1.E2M <- capHistConvert(ex1,id="fish",in.type="event",out.type="MARK") )</pre>
# convert to 'RMark' format
( ex1.E2R <- capHistConvert(ex1,id="fish",in.type="event",out.type="RMark") )</pre>
## convert converted 'individual' format ...
# to 'frequency' format (must ignore "id")
( ex1.I2F <- capHistConvert(ex1.E2I,id="fish",in.type="individual",out.type="frequency") )</pre>
# to 'MARK' format
( ex1.I2M <- capHistConvert(ex1.E2I,id="fish",in.type="individual",out.type="MARK") )</pre>
# to 'RMark' format
( ex1.I2R <- capHistConvert(ex1.E2I,id="fish",in.type="individual",out.type="RMark") )
# to 'event' format
( ex1.I2E <- capHistConvert(ex1.E2I,id="fish",in.type="individual",out.type="event") )</pre>
#' ## convert converted 'frequency' format ...
# to 'individual' format
( ex1.F2I <- capHistConvert(ex1.E2F,freq="freq",in.type="frequency") )</pre>
( ex1.F2Ia <- capHistConvert(ex1.E2F,freq="freq",in.type="frequency",include.id=TRUE) )</pre>
# to 'Mark' format
( ex1.F2M <- capHistConvert(ex1.E2F,freq="freq",in.type="frequency",out.type="MARK") )</pre>
# to 'RMark' format
( ex1.F2R <- capHistConvert(ex1.E2F,freq="freq",in.type="frequency",out.type="RMark") )</pre>
( ex1.F2Ra <- capHistConvert(ex1.E2F,freq="freq",in.type="frequency",out.type="RMark",</pre>
                              include.id=TRUE) )
# to 'event' format
( ex1.F2E <- capHistConvert(ex1.E2F, freq="freq", in.type="frequency", out.type="event") )</pre>
## convert converted 'MARK' format ...
# to 'individual' format
( ex1.M2I <- capHistConvert(ex1.E2M,freq="freq",in.type="MARK") )</pre>
( ex1.M2Ia <- capHistConvert(ex1.E2M,freq="freq",in.type="MARK",include.id=TRUE) )</pre>
# to 'frequency' format
( ex1.M2F <- capHistConvert(ex1.E2M,freq="freq",in.type="MARK",out.type="frequency") )</pre>
# to 'RMark' format
( ex1.M2R <- capHistConvert(ex1.E2M,freq="freq",in.type="MARK",out.type="RMark") )</pre>
( ex1.M2Ra <- capHistConvert(ex1.E2M,freq="freq",in.type="MARK",out.type="RMark",include.id=TRUE) )</pre>
# to 'event' format
( ex1.M2E <- capHistConvert(ex1.E2M,freq="freq",in.type="MARK",out.type="event") )</pre>
```

38 capHistConvert

```
## convert converted 'RMark' format ...
# to 'individual' format
( ex1.R2I <- capHistConvert(ex1.E2R,id="fish",in.type="RMark") )</pre>
# to 'frequency' format
( ex1.R2F <- capHistConvert(ex1.E2R,id="fish",in.type="RMark",out.type="frequency") )</pre>
# to 'MARK' format
( ex1.R2M <- capHistConvert(ex1.E2R,id="fish",in.type="RMark",out.type="MARK") )</pre>
# to 'event' format
( ex1.R2E <- capHistConvert(ex1.E2R,id="fish",in.type="RMark",out.type="event") )</pre>
## Remove semi-colon from MARK format to make a RMark 'frequency' format
ex1.E2R1 <- ex1.E2M
ex1.E2R1$freq <- as.numeric(sub(";","",ex1.E2R1$freq))</pre>
ex1.E2R1
# convert this to 'individual' format
( ex1.R2I1 <- capHistConvert(ex1.E2R1,freq="freq",in.type="RMark") )</pre>
( ex1.R2I1a <- capHistConvert(ex1.E2R1,freq="freq",in.type="RMark",include.id=TRUE) )</pre>
# convert this to 'frequency' format
( ex1.R2F1 <- capHistConvert(ex1.E2R1,freq="freq",in.type="RMark",out.type="frequency") )</pre>
# convert this to 'MARK' format
( ex1.R2M1 <- capHistConvert(ex1.E2R1,freq="freq",in.type="RMark",out.type="MARK") )</pre>
# convert this to 'event' format
( ex1.R2E1 <- capHistConvert(ex1.E2R1,freq="freq",in.type="RMark",out.type="event") )</pre>
## A small example using character ids
 ( \ ex2 <- \ data.frame(fish=c("id17","id18","id21","id17","id21","id18","id18","id19","id20"), \\
                    yr=c(1987,1987,1987,1988,1988,1989,1989,1990)) )
# convert to 'individual' format
( ex2.E2I <- capHistConvert(ex2,id="fish",in.type="event") )</pre>
# convert to 'frequency' format
( ex2.E2F <- capHistConvert(ex2,id="fish",in.type="event",out.type="frequency") )</pre>
# convert to 'MARK' format
( ex2.E2M <- capHistConvert(ex2,id="fish",in.type="event",out.type="MARK") )</pre>
# convert to 'RMark' format
( ex2.E2R <- capHistConvert(ex2,id="fish",in.type="event",out.type="RMark") )</pre>
## convert converted 'individual' format ...
# to 'frequency' format
( ex2.I2F <- capHistConvert(ex2.E2I,id="fish",in.type="individual",out.type="frequency") )</pre>
# to 'MARK' format
( ex2.I2M <- capHistConvert(ex2.E2I,id="fish",in.type="individual",out.type="MARK") )</pre>
# to 'RMark' format
( ex2.I2R <- capHistConvert(ex2.E2I,id="fish",in.type="individual",out.type="RMark") )</pre>
# to 'event' format
( ex2.I2E <- capHistConvert(ex2.E2I,id="fish",in.type="individual",out.type="event") )</pre>
## demo use of var.lbls
( ex2.E2Ia <- capHistConvert(ex2,id="fish",in.type="event",var.lbls.pre="Sample") )</pre>
( ex2.E2Ib <- capHistConvert(ex2,id="fish",in.type="event",</pre>
                             var.lbls=c("first","second","third","fourth")) )
```

capHistConvert 39

```
## demo use of event.ord
( ex2.I2Ea <- capHistConvert(ex2.E2Ib,id="fish",in.type="individual",out.type="event") )</pre>
( ex2.E2Ibad <- capHistConvert(ex2.I2Ea,id="fish",in.type="event") )</pre>
( ex2.E2Igood <- capHistConvert(ex2.I2Ea,id="fish",in.type="event",</pre>
                              event.ord=c("first","second","third","fourth")) )
## ONLY RUN IN INTERACTIVE MODE
if (interactive()) {
## A larger example of 'frequency' format (data from Rcapture package)
require(Rcapture)
data(bunting)
head(bunting)
# convert to 'individual' format
bun.F2I <- capHistConvert(bunting,in.type="frequency",freq="freq")</pre>
head(bun.F2I)
# convert to 'MARK' format
bun.F2M <- capHistConvert(bunting,id="id",in.type="frequency",freq="freq",out.type="MARK")</pre>
# convert converted 'individual' back to 'MARK' format
bun.I2M <- capHistConvert(bun.F2I,id="id",in.type="individual",out.type="MARK")</pre>
head(bun.I2M)
# convert converted 'individual' back to 'frequency' format
bun.I2F <- capHistConvert(bun.F2I,id="id",in.type="individual",out.type="frequency",</pre>
                         var.lbls.pre="Sample")
head(bun.I2F)
## A larger example of 'marked' or 'RMark' format, but with a covariate
## and when the covariate is removed there is no frequency or individual
    fish identifier.
require(marked)
data(dipper)
head(dipper)
# isolate males and females
dipperF <- subset(dipper,sex=="Female")</pre>
dipperM <- subset(dipper,sex=="Male")</pre>
# convert females to 'individual' format
dipF.R2I <- capHistConvert(dipperF,cols2ignore="sex",in.type="RMark")</pre>
head(dipF.R2I)
# convert males to 'individual' format
dipM.R2I <- capHistConvert(dipperM,cols2ignore="sex",in.type="RMark")</pre>
head(dipM.R2I)
# combine back together using combine() from gdata
require(gdata) # for combine()
dip.R2I <- combine(dipF.R2I,dipM.R2I,names=c("female","male"))</pre>
head(dip.R2I)
tail(dip.R2I)
} # end interactive
```

40 capHistSum

capHistSum

Summarize capture histories in individual fish format.

Description

Use to summarize a capture history data file that is in the "individual" fish format (see capHistConvert for a discussion of data file format types). Summarized capture history results may be used in the Lincoln-Petersent, Schnabel, Schumacher-Eschmeyer, or Jolly-Seber methods for estimating population abundance (see mrClosed and mrOpen).

Usage

```
capHistSum(df, cols2use = NULL, cols2ignore = NULL)
## S3 method for class 'CapHist'
plot(x, what = c("u", "f"), pch = 19, cex.pch = 0.7,
  lwd = 1, ...)
```

Arguments

df	A data frame that contains the capture histories (and, perhaps, other information) in "individual" fish format. See details.
cols2use	A string or numeric vector that indicates columns in df that contain the capture histories. Negative numeric values will not use those columns. Cannot use both cols2use and col2ignore. See details.
cols2ignore	A string or numeric vector that indicates columns in df that do not contain the capture histories and should be ignored. Cannot use both cols2use and col2ignore.
Х	An object from capHistSum.
what	A string that indicates what type of diagnostic plot to construct with plot. See details.
pch	A numeric that indicates the plotting character for the diagnostic plot.

capHistSum 41

cex.pch A numeric that indicates the character expansion value for the plotting characters in the diagnostic plot. The default is to be "slightly smaller" (i.e., cex.pch=0.7).

1wd A numeric that indicates the line width in the diagnostic plot.

Optional arguments to send to plot.

Details

This function requires the capture history data file to be in the "individual" fish format. See capHistConvert for a description of this (and other) formats and for methods to convert from other formats to the "individual" fish format. In addition, this function requires only the capture history portion of the data file. Thus, if df contains columns with non-capture history information (e.g., fish ID, length, location, etc.) then use cols2use= to identify which columns contain only the capture history information. Columns to use can be identified by listing the column numbers (e.g., columns 2 through 7 could be included with cols2use=2:7). In many instances it may be easier to identify columns to *exclude* which can be done by preceding the column number by a negative sign (e.g., columns 1 through 3 are excluded with cols2use=-(1:3)).

The object returned from this function can be used directly in mrClosed and mrOpen. See examples of this functionality on the help pages for those functions.

The plot function can be used to construct the two diagnostic plots described by Baillargeon and Rivest (2007). The what="f" plot will plot the log of the number of fish seen i times divided by choose(t,i) against i. The what="u" plot will plot the log of the number of fish seen for the first time on event i against i. Baillargeon and Rivest (2007) provide a table that can be used to diagnosed types of heterogeneities in capture probabilities from these plots.

Value

If the capture history data file represents only two samples, then a list with the following two components is returned.

- caphist A vector summarizing the frequency of fish with each capture history.
- sum A data.frame that contains the number of marked fish from the first sample (M), the number of captured fish in the second sample (n), and the number of recaptured (i.e. previously marked) fish in the second sample (m).

If the capture history data file represents more than two samples, then a list with the following five components is returned

- caphist A vector summarizing the frequency of fish with each capture history.
- sum A data frame that contains the the number of captured fish in the ith sample (n), the number of recaptured (i.e. previously marked) fish in the ith sample (m), the number of marked fish returned to the population following the ith sample (R; this will equal n as the function currently does not handle mortalities); the number of marked fish in the population prior to the ith sample (M); the number of fish first seen in the ith sample (u); the number of fish lasst seen in the ith sample (v); and the number of fish seen i times (f).
- methodB.top A matrix that contains the top of the Method B table used for the Jolly-Seber method (i.e., a contingency table of capture sample (columns) and last seen sample (rows)).

42 capHistSum

• methodB.bot A data.frame that contains the bottom of the Method B table used for the Jolly-Seber method (i.e., the number of marked fish in the sample (m), the number of unmarked fish in the sample (u), the total number of fish in the sample (n), and the number of marked fish returned to the population following the sample (R).

• m. array A matrix that contains the the so-called "m-array". The first column contains the number of fish captured on the ith event. The columns labelled with "cX" prefix show the number of fish originally captured in the ith row that were captured in the Xth event. The last column shows the number of fish originally captured in the ith row that were never recaptured.

IFAR Chapter

9-Abundance from Capture-Recapture Data.

Note

This function assumes that all unmarked captured fish are marked and returned to the population (i.e., no losses at the time of marking are allowed).

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL. Baillargeon, S. and Rivest, L.-P. (2007). Rcapture: Loglinear models for capture-recapture in R. Journal of Statistical Software, 19(5):1-31.

See Also

See descriptive in **Rcapture** for m. array and some of the same values in sum. See capHistConvert for a descriptions of capture history data file formats and how to convert between them. See mrClosed and mrOpen for how to estimate abundance from the summarized capture history information.

```
# data.frame with IDs in the first column
data(PikeNYPartial1)
head(PikeNYPartial1)

# Three ways to ignore first column of ID numbers
( ch1 <- capHistSum(PikeNYPartial1,cols2use=-1) )
( ch1 <- capHistSum(PikeNYPartial1,cols2ignore=1) )
( ch1 <- capHistSum(PikeNYPartial1,cols2ignore="id") )

# diagnostic plots
plot(ch1)
plot(ch1,what="f")
plot(ch1,what="u")</pre>
```

catchCurve 43

```
# An examle with only two sample events (for demonstration only)
( ch2 <- capHistSum(PikeNYPartial1,cols2use=-c(1,4:5)) )</pre>
```

catchCurve

Mortality estimates from the descending limb of a catch curve.

Description

Fits a linear model to the user-defined descending limb of a catch curve. Method functions extract estimates of the instantaneous (Z) and total annual (A) mortality rates with associated standard errors and confidence intervals. A plot method highlights the descending-limb, shows the linear model on the descending limb, and, optionally, prints the estimated Z and A.

Usage

```
catchCurve(x, ...)
## Default S3 method:
catchCurve(x, catch, ages2use = age, weighted = FALSE,
  ...)
## S3 method for class 'formula'
catchCurve(x, data, ages2use = age, weighted = FALSE, ...)
## S3 method for class 'catchCurve'
summary(object, type = c("params", "lm"), ...)
## S3 method for class 'catchCurve'
coef(object, type = c("params", "lm"), ...)
## S3 method for class 'catchCurve'
anova(object, ...)
## S3 method for class 'catchCurve'
confint(object, parm = c("all", "both", "Z", "A"),
  level = conf.level, conf.level = 0.95, type = c("params", "lm"), ...)
## S3 method for class 'catchCurve'
plot(x, pos.est = "topright", cex.est = 0.95,
  ylab = "log(Catch)", xlab = "Age", col.pt = "gray30",
  col.mdl = "black", lwd = 2, lty = 1, ...)
```

Arguments

Х

A numerical vector of the assigned ages in the catch curve or a formula of the form catch~age when used in catchCurve. An object saved from catchCurve (i.e., of class catchCurve) when used in the methods.

44 catchCurve

catch	A numerical vector of the catches or CPUEs for the ages in the catch curve. Not used if x is a formula.
ages2use	A numerical vector of the ages that define the descending limb of the catch curve.
weighted	A logical that indicates whether a weighted regression should be used. See details.
data	A data frame from which the variables in the x formula can be found. Not used if x is not a formula.
object	An object saved from the catchCurve call (i.e., of class catchCurve).
type	A string that indicates what type of summary should be returned. If type="lm" then summaries of the underlying linear model are returned. If type="params" then summaries of the Z and A parameters are returned.
parm	A numeric or string (of parameter names) vector that specifies which parameters are to be given confidence intervals If missing, all parameters are considered.
level	Same as conf.level. Used for compatability with the generic confint function.
conf.level	A number representing the level of confidence to use for constructing confidence intervals.
pos.est	A string to identify where to place the estimated mortality rates on the plot. Can be set to one of "bottomright", "bottom", "bottomleft", "left", "topleft", "top", "topright", "right" or "center" for positioning the estimated mortality rates on the plot. Typically "bottomleft" (DEFAULT) and "topright" will be "out-of-the-way" placements. Set pos.est to NULL to remove the estimated mortality rates from the plot.
cex.est	A single numeric characther expansion value for the estimated mortaliry rates on the plot.
ylab	A label for the y-axis ("log(Catch)" is the default).
xlab	A label for the x-axis ("Age" is the default).
col.pt	A string that indicates the color of the plotted points.
col.mdl	A string that indicates the color of the fitted line.
lwd	A numeric that indicates the line width of the fitted line.
lty	A numeric that indicates the type of line used for the fitted line.
	Additional arguments for methods.

Details

The default is to use all ages in the age vector. This is appropriate only when the age and catch vectors contain only the ages and catches on the descending limb of the catch curve. Use ages2use to isolate only the catch and ages on the descending limb.

If weighted=TRUE then a weighted regression is used where the weights are the log(number) at each age predicted from the unweighted regression of log(number) on age (as proposed by Maceina and Bettoli (1998)).

catchCurve 45

Value

A list that contains the following items:

- age The original vector of assigned ages.
- catch The original vector of observed catches or CPUEs.
- age.e A vector of assigned ages for which the catch curve was fit.
- log.catch.e A vector of log catches or CPUEs for which the catch curve was fit.
- W A vector of weights used in the catch curve fit. Will be NULL unless weighted=TRUE.
- Im An 1m object from the fit to the ages and log catches or CPUEs on the descending limb (i.e., in age.e and log.catch.e).

Testing

Tested the results of catch curve, both unweighted and weighted, against the results in Miranda and Bettoli (2007). Results for Z and the SE of Z matched perfectly. Tested the unweighted results against the results from agesurv in **fishmethods** using the rockbass data.frame in **fishmethods**. Results for Z and the SE of Z matched perfectly.

IFAR Chapter

11-Mortality.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Maceina, M.J., and P.W. Bettoli. 1998. Variation in largemouth bass recruitment in four mainstream impoundments on the Tennessee River. North American Journal of Fisheries Management 18:998-1003.

Ricker, W.E. 1975. Computation and interpretation of biological statistics of fish populations. Technical Report Bulletin 191, Bulletin of the Fisheries Research Board of Canada.

See Also

See agesurv in **fishmethods** for similar functionality. See chapmanRobson and agesurvcl in **fishmethods** for alternative methods to estimate mortality rates. See metaM for empirical methods to estimate natural mortality.

```
data(BrookTroutTH)
plot(catch~age,data=BrookTroutTH,pch=19)

## demonstration of formula notation
cc1 <- catchCurve(catch~age,data=BrookTroutTH,ages2use=2:6)</pre>
```

```
summary(cc1)
plot(cc1)
coef(cc1)
confint(cc1)
## demonstration of excluding ages2use
cc2 <- catchCurve(catch~age,data=BrookTroutTH,ages2use=-c(0,1))</pre>
summary(cc2)
plot(cc2)
## demonstration of using weights
cc3 <- catchCurve(catch~age,data=BrookTroutTH,ages2use=2:6,weighted=TRUE)</pre>
summary(cc3)
plot(cc3)
## demonstration of returning the linear model results
summary(cc3,type="lm")
coef(cc3,type="lm")
confint(cc3,type="lm")
## demonstration of ability to work with missing age classes
df <- data.frame(age=c( 2, 3, 4, 5, 7, 9,12),</pre>
                 ct= c(100,92,83,71,56,35, 1))
cc4 <- catchCurve(ct~age,data=df,ages2use=4:12)</pre>
summary(cc4)
plot(cc4)
## demonstration of ability to work with missing age classes
## evein if catches are recorded as NAs
df <- data.frame(age=c( 2, 3, 4, 5, 6, 7, 8, 9,10,11,12),
                 ct= c(100,92,83,71,NA,56,NA,35,NA,NA, 1))
cc5 <- catchCurve(ct~age,data=df,ages2use=4:12)</pre>
summary(cc5)
plot(cc5)
```

chapmanRobson

Computes Chapman-Robson estimates of S and Z.

Description

Computes the Chapman-Robson estimates of annual survival rate (S) and instantaneous mortality rate (Z) from catch-at-age data on the descending limb of a catch-curve. Method functions extract estimates with associated standard errors and confidence intervals. A plot method highlights the descending-limb, shows the linear model on the descending limb, and, optionally, prints the estimated Z and A.

Usage

```
chapmanRobson(x, ...)
```

```
## Default S3 method:
chapmanRobson(x, catch, ages2use = age,
    zmethod = c("Smithetal", "Hoenigetal", "original"), ...)

## S3 method for class 'formula'
chapmanRobson(x, data, ages2use = age,
    zmethod = c("Smithetal", "Hoenigetal", "original"), ...)

## S3 method for class 'chapmanRobson'
summary(object, verbose = FALSE, ...)

## S3 method for class 'chapmanRobson'
confint(object, parm = c("all", "both", "S", "Z"),
    level = conf.level, conf.level = 0.95, ...)

## S3 method for class 'chapmanRobson'
plot(x, pos.est = "topright", cex.est = 0.95,
    ylab = "Catch", xlab = "Age", col.pt = "gray30", ...)
```

Arguments

х	A numerical vector of the assigned ages in the catch curve or a formula of the form catch~age when used in chapmanRobson. An object saved from chapmanRobson (i.e., of class chapmanRobson) when used in the methods.
catch	A numerical vector of the catches or CPUEs for the ages in the catch curve. Not used if x is a formula.
ages2use	A numerical vector of the ages that define the descending limb of the catch curve.
zmethod	A string that indicates the method to use for estimating Z. See details.
data	A data frame from which the variables in the \boldsymbol{x} formula can be found. Not used if \boldsymbol{x} is not a formula.
object	An object saved from the chapmanRobson call (i.e., of class chapmanRobson).
verbose	A logical that indicates whether the method should return just the estimate (FALSE; default) or a more verbose statement.
parm	A numeric or string (of parameter names) vector that specifies which parameters are to be given confidence intervals If missing, all parameters are considered.
level	Same as conf.level. Used for compatability with the generic confint function.
conf.level	A number representing the level of confidence to use for constructing confidence intervals.
pos.est	A string to identify where to place the estimated mortality rates on the plot. Can be set to one of "bottomright", "bottom", "bottomleft", "left", "topleft", "top", "topright", "right" or "center" for positioning the estimated mortality rates on the plot. Typically "bottomleft" (DEFAULT) and "topright"

mated mortality rates from the plot.

will be "out-of-the-way" placements. Set pos.est to NULL to remove the esti-

cex.est	A single numeric characther expansion value for the estimated mortaliry rates on the plot.
ylab	A label for the y-axis ("Catch" is the default).
xlab	A label for the x-axis ("Age" is the default).
col.pt	a string that indicates the color of the plotted points.
	Additional arguments for methods.

Details

The default is to use all ages in the age vector. This is only appropriate if the age and catch vectors contain only the ages and catches on the descending limb of the catch curve. Use ages2use to isolate only the catch and ages on the descending limb.

The Chapman-Robson method provides an estimate of the annual survival rate, with the annual mortality rate (A) determined by 1-S. The instantaneous mortality rate is often computed as -log(S). However, Hoenig et al. (1983) showed that this produced a biased (over)estimate of Z and provided a correction. The correction is applied by setting zmethod="Hoenigetal". Smith et al. (2012) showed that the Hoenig et al. method should be corrected for a variance inflation factor. This correction is applied by setting zmethod="Smithetal" (which is the default behavior). Choose zmethod="original" to use the original estimates for Z and it's SE as provided by Chapman and Robson.

Value

A list with the following items:

- · age the original vector of assigned ages.
- catch the original vector of observed catches or CPUEs.
- age.e a vector of assigned ages used to estimate mortalities.
- catch.e a vector of catches or CPUEs used to estimate mortalities.
- age.r a vector of recoded ages used to estimate mortalities. See references.
- n a numeric holding the intermediate calcualtion of n. See references.
- T a numeric holding the intermediate calcualtion of T. See references.
- est A 2x2 matrix that contains the estimates and standard errors for S and Z.

Testing

Tested the results of chapmanRobson against the results in Miranda and Bettoli (2007). The point estimates of S matched perfectly but the SE of S did not because Miranda and Bettoli used a rounded estimate of S in the calculation of the SE of S but chapmanRobson does not.

Tested the results against the results from agesurv in **fishmethods** using the rockbass data.frame in **fishmethods**. Results for Z and the SE of Z matched perfectly for non-bias-corrected results. The estimate of Z, but not the SE of Z, matched for the bias-corrected (following Smith et al. (2012)) results. **FSA** uses equation 2 from Smith et al. (2012) whereas **fishmethods** appears to use equation 5 from the same source to estimate the SE of Z.

IFAR Chapter

```
11-Mortality.
```

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Chapman, D.G. and D.S. Robson. 1960. The analysis of a catch curve. Biometrics. 16:354-368.

Hoenig, J.M. and W.D. Lawing, and N.A. Hoenig. 1983. Using mean age, mean length and median length data to estimate the total mortality rate. International Council for the Exploration of the Sea, CM 1983/D:23, Copenhagen.

Ricker, W.E. 1975. Computation and interpretation of biological statistics of fish populations. Technical Report Bulletin 191, Bulletin of the Fisheries Research Board of Canada.

Robson, D.S. and D.G. Chapman. 1961. Catch curves and mortality rates. Transactions of the American Fisheries Society. 90:181-189.

Smith, M.W., A.Y. Then, C. Wor, G. Ralph, K.H. Pollock, and J.M. Hoenig. 2012. Recommendations for catch-curve analysis. North American Journal of Fisheries Management. 32:956-967.

See Also

See agesurv in **fishmethods** for similar functionality. See catchCurve and agesurvcl in **fishmethods** for alternative methods. See metaM for empirical methods to estimate natural mortality.

```
data(BrookTroutTH)
plot(catch~age,data=BrookTroutTH,pch=19)
## demonstration of formula notation
cr1 <- chapmanRobson(catch~age,data=BrookTroutTH,ages2use=2:6)</pre>
summary(cr1)
summary(cr1,verbose=TRUE)
confint(cr1)
plot(cr1)
## demonstration of excluding ages2use
cr2 <- chapmanRobson(catch~age,data=BrookTroutTH,ages2use=-c(0,1))</pre>
summary(cr2)
plot(cr2)
## demonstration of ability to work with missing age classes
age < c( 2, 3, 4, 5, 7, 9,12)
ct < c(100,92,83,71,56,35, 1)
cr3 <- chapmanRobson(age,ct,4:12)</pre>
summary(cr3)
plot(cr3)
```

50 chooseColors

gentina.	ChinookArg	Lengths and weights of Chinook Salmon from three locations in Argentina.
----------	------------	--

Description

Lengths and weights of Chinook Salmon from three locations in Argentina.

Format

A data frame with 112 observations on the following 3 variables:

```
tl Total length (cm)
```

w Weight (kg)

loc Capture location (Argentina, Petrohue, Puyehue)

Topic(s)

• Length-weight

Examples

```
data(ChinookArg)
str(ChinookArg)
head(ChinookArg)
op <- par(mfrow=c(2,2),pch=19)
plot(w~tl,data=ChinookArg,subset=loc=="Argentina")
plot(w~tl,data=ChinookArg,subset=loc=="Petrohue")
plot(w~tl,data=ChinookArg,subset=loc=="Puyehue")
par(op)</pre>
```

chooseColors

Create a list of colors from among a variety of color palettes.

Description

Create a list of colors from among a variety of color palettes.

Usage

```
chooseColors(pal = paletteChoices(), num, ...)
paletteChoices()
```

CodNorwegian 51

Arguments

pal	A character that is the name of a palette. Must be one of "rich", "cm", "default",
	"grey", "gray", "heat", "jet", "rainbow", "topo", or "terrain", which are given in
	paletteChoices.
num	The number of colors to be returned.
	Other arguments to the various palette functions.

Value

A vector of colors of length num.

Author(s)

```
Derek H. Ogle, <derek@derekogle.com>
```

See Also

See rich.colors in **gplots**, cm.colors, heat.colors, topo.colors, terrain.colors, rainbow, colorRampPalette, and colors.

Examples

```
n <- 20
# Color Wheels
pie(rep(1,n), col=chooseColors("rich",n))
pie(rep(1,n), col=chooseColors("rainbow",n))
pie(rep(1,n), col=chooseColors("topo",n))
pie(rep(1,n), col=chooseColors("gray",n))
pie(rep(1,n), col=chooseColors("jet",n))</pre>
```

CodNorwegian

Stock and recruitment data for Norwegian cod, 1937-1960.

Description

Norwegian cod (Gadus morhua) stock and recruitment by year, 1937-1960.

Format

A data frame of 24 observations on the following 3 variables:

```
year Year of data.recruits Recruits – year-class strength index.stock Spawning stock index.
```

Topic(s)

• Stock-recruit

52 col2rgbt

Source

From Garrod, D.J. 1967. Population dynamics of the Arcto-Norwegian cod. Journal of the Fisheries Research Board of Canada, 24:145-190.

Examples

```
data(CodNorwegian)
str(CodNorwegian)
head(CodNorwegian)
op <- par(mfrow=c(1,2))</pre>
plot(recruits~year,data=CodNorwegian,type="l")
plot(recruits~stock,data=CodNorwegian)
par(op)
```

col2rgbt

Converts an R color to RGB (red/green/blue) including a transparency (alpha channel).

Description

Converts an R color to RGB (red/green/blue) including a transparency (alpha channel). Similar to col2rgb except that a transparncy (alpha channel) can be included.

Usage

```
col2rgbt(col, transp = 1)
```

Arguments

col

A vector of any of the three kinds of R color specifications (i.e., either a color name (as listed by colors()), a hexadecimal string of the form "#rrggbb" or "#rrggbbaa" (see rgb), or a positive integer i meaning palette()[i].

transp

A numeric vector that indicates the transparency level for the color. The transpaency values must be greater than 0. Transparency values greater than 1 are interpreted as the number of points plotted on top of each other before the transparency is lost and is, thus, transformed to the inverse of the transparency value provided.

Value

A vector of hexadecimal strings of the form "#rrggbbaa" as would be returned by rgb.

Author(s)

```
Derek H. Ogle, <derek@derekogle.com>
```

See Also

See col2rgb for similar functionality.

compIntercepts 53

Examples

```
col2rgbt("black")
col2rgbt("black",1/4)
clrs <- c("black","blue","red","green")
col2rgbt(clrs)
col2rgbt(clrs,1/4)
trans <- (1:4)/5
col2rgbt(clrs,trans)</pre>
```

compIntercepts

Tests for significant differences among all pairs of intercepts in a dummy variable regression.

Description

Tests for significant differences among all pairs of intercepts in a dummy variable regression where the dummy variables all stem from one factor.

Usage

```
compIntercepts(mdl, common.cov = mean(x), conf.level = 0.95,
   digits = getOption("digits"), ...)
## S3 method for class 'compIntercepts'
print(x, ...)
```

Arguments

mdl A	A 1m object.
	A value to be used as the common value of the covariate in the adjustment pro- ess. See details.
	a single number that represents the level of confidence to use for constructing onfidence intervals.
digits A	a numeric that controls the number of digits to print.
x A	a compIntercepts object (i.e., returned from compIntercepts).
0	Other arguments to be passed to the TukeyHSD or print functions.

Details

In a dummy variable regression without the interaction(s) between the covariate (x) and the dummy variable(s) (i.e., parallel lines) the coefficient for the dummy variables tests for a difference in intercepts between the level of the dummy variable and the reference level. Thus, all dummy variables from a particular linear model fit only compare intercepts with the reference level. Other intercept comparisons may be found by changing the reference level, which requires refitting the model.

Alternatively, Tukey's HSD method of multiple comparisons may be used, but this requires adjusting the original observations as if the original observations were all collected at the exact same value

54 compIntercepts

of the covariate (x). Because of this required adjustment, the TukeyHSD function is inappropriate for testing for difference in intercepts in a dummy variable regression.

This function provides a statistical comparison of all pairs of intercepts by first adjusting the observed data to a common value of the covariate (common.cov), computing a one-way ANOVA to determine if the mean adjusted values differ by level of the group factor in the original dummy variable regression, and then submitting the one-way ANOVA results to the TukeyHSD function to determine for which levels the mean adjusted values differ. The levels for which the mean adjusted values differ are also the levels for which the intercepts differ.

The default is to compute the adjusted values at the mean value of the covariate (i.e., common. cov=mean(x). However, if interest is in the intercepts (i.e., at X=0) then common. cov=0 should be used instead.

Value

A list with the following four components:

comparison comparison results as returned from TukeyHSD.

The value of the common covariate sent in common.cov.

A vector of values of the repsonse variable adjusted to the common.cov value of the covariate. This vector can A vector of mean adjusted values at the value of the common covariate.

The value sent in digits.

The name of the response (LHS) variable.

The name of the covariate variable.

The print function prints the comparison and adjusted means in a nice format.

Author(s)

```
Derek H. Ogle, <derek@derekogle.com>
```

See Also

TukeyHSD and compSlopes from **FSA**.

```
## Reduct Mirex data to years where slopes don't differ to
## illustrate this function ... see compSlopes() for
## analysis of full data set.
data(Mirex)
Mirex <- Mirex[Mirex$year!="1996" & Mirex$year!="1999",]
Mirex$year <- factor(Mirex$year)
## Fit DVR, see that slopes don't differ,
## compare intercepts, visualize results
lm1 <- lm(mirex~weight*year,data=Mirex)
anova(lm1)
compIntercepts(lm1)
## Fit model without interaction to avoid warning, but
## note that the compIntercepts() results are the same
lm2 <- lm(mirex~weight+year,data=Mirex)</pre>
```

compSlopes 55

```
compIntercepts(lm2)
fitPlot(lm1,legend="topleft")
```

compSlopes Tests for significant differences among all pairs of slopes in a dummy variable regression (DVR).

Description

Tests for significant differences among all pairs of slopes in a dummy variable regression where the dummy variables all stem from one factor.

Usage

```
compSlopes(mdl, method = stats::p.adjust.methods, conf.level = 0.95,
  order.slopes = TRUE, digits = getOption("digits"))

## S3 method for class 'compSlopes'
print(x, ...)
```

Arguments

mdl	A 1m object.
method	A string indicating the method of p-value adjustment to use. See details and ${\tt p.adjust.methods}.$
conf.level	A single number that represents the level of confidence to use for constructing confidence intervals.
order.slopes	A logical indicating whether the slopes should be ordered from smallest to largest upon output.
digits	A numeric that controls the number of digits to print.
x	A compSlopes object (i.e., returned from compSlopes).
	Other arguments sent to print.

Details

In a dummy variable regression the coefficient for the interaction between the covariate (x) and a dummy variable tests for a difference in slopes between the level of the dummy variable and the reference level. Thus, all dummy variables from a particular linear model fit only compare slopes with the reference level. Other slope comparisons may be found by changing the reference level, which requires refitting the model. This function automates this sequential process and produces a data.frame that shows the estimated difference, an unadjusted confidence interval for the difference, and the unadjusted and adjusted (for multiple comparisons) p-values for testing that the difference in slopes is equal to zero for each pair of levels. The adjusted p-values may be computed with any of the methods coded in p.adjust (see p.adjust.methods).

56 compSlopes

Value

A list with three components. The first component contains the p-value adjustment method in method. The second component, called comparisons, is a data frame that contains the following:

comparison	Description of how the difference in levels was computed.
diff	The estimated difference in slope values.
lwr	Lower confidence bound for difference in slope values.
upr	Upper confidence bound for difference in slope values.
p.unadj	Unadjusted p-value for testing the difference in slopes is zero
p.adi	Adjusted p-value for testing the difference in slopes is zero.

The third component, called slopes, is a data frame that contains the following:

level	A level name.
slope	The estimated slope value for the given level.
XX LCI	Lower confidence bound for difference in slope values.
XX UCI	Upper confidence bound for difference in slope values.
p.unadj	Unadjusted p-value for testing the slope is zero.
p.adj	Adjusted p-value for testing the slope is zero.

The print function prints the results nicely.

Note

This function only works for linear models with one factor variable.

Author(s)

```
Derek H. Ogle, <derek@derekogle.com>
```

See Also

```
compIntercepts in FSA.
```

```
data(Mirex)
Mirex$year <- factor(Mirex$year)
# fit a dummy variable regression, see that slopes differ
lm1 <- lm(mirex~weight*year,data=Mirex)
anova(lm1)
# compare all pairs of slopes using default Holm control
compSlopes(lm1)
# compare all pairs of slopes using the false discovery rate control
compSlopes(lm1,method="fdr")
# visualize the results
fitPlot(lm1)</pre>
```

confint.nlsBoot 57

confint.nlsBoot	Associated \$3 methods:	for nlsBoot from nlstools.
COIII III C. III SDOOC	11550ctatea 55 memous	or misboot from mistoots.

Description

Provides S3 methods to construct non-parametric bootstrap confidence intervals and hypothesis tests for parameter values and predicted values of the response variable for a nlsBoot object from the **nlstools** package.

Usage

```
## S3 method for class 'nlsBoot'
confint(object, parm = NULL, level = conf.level,
    conf.level = 0.95, plot = FALSE, err.col = "black", err.lwd = 2,
    rows = NULL, cols = NULL, ...)

## S3 method for class 'nlsBoot'
predict(object, FUN, MARGIN = 1, conf.level = 0.95,
    digits = NULL, ...)

htest(object, ...)

## S3 method for class 'nlsBoot'
htest(object, parm = NULL, bo = 0, alt = c("two.sided",
    "less", "greater"), plot = FALSE, ...)
```

Arguments

object	An object saved from nlsBoot().
parm	An integer that indicates which parameter to compute the confidence interval or hypothesis test for. The confidence interval Will be computed for all parameters if NULL.
level	Same as conf.level. Used for compatability with the main confint.
conf.level	A level of confidence as a proportion.
plot	A logical that indicates whether a plot should be constructed. If confint, then a histogram of the parm parameters from the bootstrap samples with error bars that illustrate the bootstrapped confidence intervals will be constructed. If codehtest, then a histogram of the parm parameters with a vertical lines illustrating the bovalue will be constructed.
err.col	A single numeric or character that identifies the color for the error bars on the plot.
err.lwd	A single numeric that identifies the line width for the error bars on the plot.
rows	A numeric that contains the number of rows to use on the graphic.
cols	A numeric that contains the number of columns to use on the graphic.

58 confint.nlsBoot

FUN	The function to be applied for the prediction. See the examples.
MARGIN	A single numeric that indicates the margin over which FUN is applied. MARGIN=1 will apply to each row and is the default.
digits	A single numeric that indicates the number of digits for the result.
bo	The null hypothesized parameter value.
alt	A string that identifies the "direction" of the alternative hypothesis. See details.
	Additional arguments to functions.

Details

confint finds the two quantiles that have the proportion (1-conf.level)/2 of the bootstrapped parameter estimates below and above. This is an approximate 100conf.level% confidence interval.

In htest the "direction" of the alternative hypothesis is identified by a string in the alt= argument. The strings may be "less" for a "less than" alternative, "greater" for a "greater than" alternative, or "two.sided" for a "not equals" alternative (the DEFAULT). In the one-tailed alternatives the p-value is the proportion of bootstrapped parameter estimates in object\$coefboot that are extreme of the null hypothesized parameter value in bo. In the two-tailed alternative the p-value is twice the smallest of the proportion of bootstrapped parameter estimates above or below the null hypothesized parameter value in bo.

In predict, a user-supplied function is applied to each row of the coefBoot object in a nlsBoot object and then finds the median and the two quantiles that have the proportion (1-conf.level)/2 of the bootstrapped predictions below and above. The median is returned as the predicted value and the quantiles are returned as an approximate 100conf.level% confidence interval for that prediction.

Value

confint returns a matrix with as many rows as columns (i.e., parameter estimates) in the object\$coefboot data frame and two columns of the quantiles that correspond to the approximate confidence interval.

htest returns a matrix with two columns. The first column contains the hypothesized value sent to this function and the second column is the corresponding p-value.

predict returns a matrix with one row and three columns, with the first column holding the predicted value (i.e., the median prediction) and the last two columns holding the approximate confidence interval.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

See Also

See summary.nlsBoot in nlstools

CutthroatAL 59

Examples

```
data(Ecoli)
fnx <- function(days,B1,B2,B3) {</pre>
  if (length(B1) > 1) {
    B2 <- B1[2]
    B3 <- B1[3]
    B1 <- B1[1]
  B1/(1+exp(B2+B3*days))
nl1 <- nls(cells~fnx(days,B1,B2,B3),data=Ecoli,start=list(B1=6,B2=7.2,B3=-1.45))</pre>
if (require(nlstools)) {
  nl1.boot <- nlstools::nlsBoot(nl1,niter=99) # way too few
  confint(nl1.boot, "B1")
  confint(nl1.boot,c(2,3))
  confint(nl1.boot,conf.level=0.90)
  predict(nl1.boot,fnx,days=3)
  htest(nl1.boot,1,bo=6,alt="less")
}
```

CutthroatAL

Capture histories (9 samples) of Cutthroat Trout from Auke Lake.

Description

Individual capture histories of Cutthroat Trout (*Oncorhynchus clarki*) in Auke Lake, Alaska, from samples taken in 1998-2006.

Format

A data frame with 1684 observations on the following 10 variables.

id Unique identification numbers for each fish.

y1998 Indicator variable for whether the fish was captured in 1998 (1=captured).

y1999 Indicator variable for whether the fish was captured in 1999 (1=captured).

y2000 Indicator variable for whether the fish was captured in 2000 (1=captured).

y2001 Indicator variable for whether the fish was captured in 2001 (1=captured).

y2002 Indicator variable for whether the fish was captured in 2002 (1=captured).

y2003 Indicator variable for whether the fish was captured in 2003 (1=captured). y2004 Indicator variable for whether the fish was captured in 2004 (1=captured).

•

y2005 Indicator variable for whether the fish was captured in 2005 (1=captured).

y2006 Indicator variable for whether the fish was captured in 2006 (1=captured).

depletion depletion

Topic(s)

- · Population size
- Abundance
- · Mark-recapture
- · Jolly-Seber method
- · Capture history

Source

Entered from Appendix A.3 of Harding, R.D., C.L. Hoover, and R.P. Marshall. 2010. Abundance of Cutthroat Trout in Auke Lake, Southeast Alaska, in 2005 and 2006. Alaska Department of Fish and Game Fisheries Data Series No. 10-82 into "RMark" format (see CutthroatALf in FSAdata) and then converted to individual format with capHistConvert.

See Also

See mrOpen for an example analysis.

Examples

```
data(CutthroatAL)
str(CutthroatAL)
head(CutthroatAL)
```

depletion

Computes the Leslie or Delury population estimate from catch and effort data.

Description

Computes the Leslie or Delury estimates of population size and catchability coefficient from paired catch and effort data. The Ricker modification may also be used.

Usage

```
depletion(catch, effort, method = c("Leslie", "DeLury", "Delury"),
   Ricker.mod = FALSE)

## S3 method for class 'depletion'
summary(object, type = c("params", "lm"),
   verbose = FALSE, digits = getOption("digits"), ...)

## S3 method for class 'depletion'
coef(object, type = c("params", "lm"),
   digits = getOption("digits"), ...)
```

depletion 61

```
## S3 method for class 'depletion'
confint(object, parm = c("No", "q", "lm"),
  level = conf.level, conf.level = 0.95, digits = getOption("digits"),
    ...)

## S3 method for class 'depletion'
anova(object, ...)

## S3 method for class 'depletion'
plot(x, xlab = NULL, ylab = NULL, pch = 19,
    col.pt = "black", col.mdl = "gray70", lwd = 1, lty = 1,
    pos.est = "topright", cex.est = 0.95, ...)
```

Arguments

lty

catch	A numeric vector of catches of fish at each time.
effort	A numeric vector of efforts expended at each time.
method	A single string that indicates which depletion method to use
Ricker.mod	A single logical that indicates whether to use the modification proposed by Ricker (=TRUE) or not (=FALSE, default).
object	An object saved from the removal call (i.e., of class depletion).
type	A string that indicates the type of summary or coefficients to extract. If type="params" (the default) then results for No and q are returned. If type="lm" then results for the underlying linear model are returned.
verbose	A logical that indicates whether a reminder of the method used should be printed with the summary results.
digits	A single numeric indicating the number of digits to round the output.
parm	A specification of which parameters are to be given confidence intervals, either a vector of numbers or a vector of names. If missing, all parameters are considered.
level	Same as conf.level but used for compatability with generic confint function.
conf.level	A single number that represents the level of confidence to use for constructing confidence intervals.
X	An object saved from the depletion call (i.e., of class depletion).
xlab	A label for the x-axis.
ylab	A label for the y-axis.
pch	A numeric that indicates the type of plotting character.
col.pt	A string that indicates the color of the plotted points.
col.mdl	A string that indicates the color of the fitted line.
lwd	A numeric that indicates the line width of the fitted line.

A numeric that indicates the type of line used for the fitted line.

62 depletion

A single string to identify where to place the estimated population estimate and catchability on the plot. Can be set to one of "bottomright", "bottom", "bottomleft", "left", "topleft", "top", "topright", "right" or "center" for positioning the estimated mortality rates on the plot. Typically "bottomleft" (DEFAULT) and "topright" will be "out-of-the-way" placements. Set pos.est to NULL to remove the estimated population size and catchability coefficient from the plot.

cex.est

A single numeric that identifies the character expansion value for the estimated population estimate and catchability placed on the plot.

... Additional arguments for methods.

Details

For the Leslie method, a linear regression model of catch-per-unit-effort on cumulative catch prior to the sample is fit. The catchability coefficient (q) is estimated from the negative of the slope and the initial population size (No) is estimated by dividing the intercept by the catchability coefficient. If Ricker.mod=TRUE then the cumulative catch is modified to be the cumulative catch prior to the sample plus half of the catch of the current sample.

For the DeLury method, a linear regression model of log (catch-per-unit-effort) on cumulative effort is fit. The catchability coefficient (q) is estimated from the negative of the slope and the initial population size (No) is estimated by dividing the intercept as an exponent of e by the catchability coefficient. If Ricker.mod=TRUE then the cumulative effort is modified to be the cumulative effort prior to the sample plus half of the effort of the current sample.

Standard errors for the catchability and population size estimates are computed from formulas on page 298 (for Leslie) and 303 (for DeLury) from Seber (2002). Confidence intervals are computed using standard large-sample normal distribution theory with the regression error df.

Value

A list with the following items:

- method A string that indicates whether the "Leslie" or "Delury" model was used.
- catch The original vector of catches.
- effort The orignal vector of efforts.
- cpe A computed vector of catch-per-unit-effort for each time.
- KorE A computed vector of cumulative catch (K; Leslie method) or effort (E; Delury method).
- Im The 1m object from the fit of CPE on K (Leslie method) or log(CPE) on E (Delury method).
- est A 2x2 matrix that contains the estimates and standard errors for No and q.

testing

The Leslie method without the Ricker modification and the DeLury method with the Ricker modification matches the results from deplet in **fishmethods** for the darter (from **fishmethods**), LobsterPEI and BlueCrab from **FSAdata**, and SMBassLS for N0 to whole numbers, the SE for No to one decimal, q to seven decimals, and the SE of q to at least five decimals.

depletion 63

The Leslie method matches the results of Seber (2002) for N0, q, and the CI for Q but not the CI for N (which was so far off that it might be that Seber's result is incorrect) for the lobster data and the q and CI for q but the NO or its CI (likely due to lots of rounding in Seber 2002) for the Blue Crab data.

The Leslie and DeLury methods match the results of Ricker (1975) for No and Q but not for the CI of No (Ricker used a very different method to compute CIs).

IFAR Chapter

10-Abundance from Depletion Data.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL. Ricker, W.E. 1975. Computation and interpretation of biological statistics of fish populations. Technical Report Bulletin 191, Bulletin of the Fisheries Research Board of Canada.

Seber, G.A.F. 2002. The Estimation of Animal Abundance. Edward Arnold, Second edition (reprinted).

See Also

See removal for related functionality and deplet in fishmethods for similar functionality.

```
data(SMBassLS)
## Leslie model examples
# no Ricker modification
11 <- depletion(SMBassLS$catch,SMBassLS$effort,method="Leslie")</pre>
coef(11)
summary(11)
summary(11,verbose=TRUE)
confint(11)
summary(11,type="lm")
plot(11)
# with Ricker modification
12 <- depletion(SMBassLS$catch,SMBassLS$effort,method="Leslie",Ricker.mod=TRUE)
summary(12)
confint(12)
plot(12)
## Delury model examples
# no Ricker modification
d1 <- depletion(SMBassLS$catch,SMBassLS$effort,method="Delury")</pre>
```

64 dunnTest

```
coef(d1)
summary(d1)
summary(d1,verbose=TRUE)
confint(d1)
summary(d1,type="lm")
plot(d1)

# with Ricker modification
d2 <- depletion(SMBassLS$catch,SMBassLS$effort,method="Delury",Ricker.mod=TRUE)
summary(d2)
confint(d2)
plot(d2)</pre>
```

dunnTest

Dunn's Kruskal-Wallis Multiple Comparisons.

Description

Performs Dunn's (1964) test of multiple comparisons following a significant Kruskal-Wallis test, possibly with a correction to control the experimentwise error rate.

Usage

```
dunnTest(x, ...)
## Default S3 method:
dunnTest(x, g, method = dunn.test::p.adjustment.methods[c(4,
    2:3, 5:8, 1)], two.sided = TRUE, ...)
## S3 method for class 'formula'
dunnTest(x, data = NULL,
    method = dunn.test::p.adjustment.methods[c(4, 2:3, 5:8, 1)],
    two.sided = TRUE, ...)
## S3 method for class 'dunnTest'
print(x, dunn.test.results = FALSE, ...)
```

Arguments

Х	A numeric vector of data values or a formula of the form x~g.
g	A factor vector or a (non-numeric) vector that can be coerced to a factor vector.
method	A single string that identifies the method to use to control the experimentwise error rate. See the list of methods in p.adjustment.methods.
two.sided	A single logical that indicates whether a two-sided p-value should be returned (TRUE; default) or not. See details.
data	A data.frame that minimally contains x and g.

dunnTest 65

dunn.test.results

A single logical that indicates whether the results that would have been printed by dunn.test function in dunn.test are shown.

... Not yet used.

Details

This function performs "Dunn's" test of multiple comparisons following a Kruskal-Wallis test. Unadjusted two-sided p-values for each pairwise comparison among groups are computed following Dunn's description as implemented in the dunn. test function from dunn.test. These p-values may be adjusted using the methods in the p.adjustment.methods function in dunn.test as implemented in the dunn.test function in dunn.test.

This function is largely a wrapper for the dunn.test function in dunn.test. Changes here are the possible use of formula notation, results that are not printed by the main function (but are printed in a more useful format (in my opinion) by the print function), the p-values are adjusted by default with the "holm" method, and two-sided p-values (two times the p-value produced by the dunn.test function) are returned with two.sided=TRUE. This function returns the same results (in a different format) as the dunn.test function from dunn.test when two.sided=FALSE. See dunn.test function in dunn.test for a more details underlying these computations.

Value

A list with three items – method is the long name of the method used to control the experimentwise error rate, dtres is the strings that would have been printed by dunn. test function in dunn.test, and a data.frame in res with the following variables:

- Comparison Labels for each pairwise comparison.
- Z Values for the Z test statistic for each comparison.
- P.unadj Unadjusted p-values for each comparison.
- P.adj Adjusted p-values for each comparison.

Note

The data frame will be reduced to only those rows that are complete cases for x and g. In other words, rows with missing data for either x or g are removed from the analysis.

There are a number of functions in other packages that purport to do similar analyses.

The pairw.kw function from the **asbio** package performs the Dunn test with the Bonferroni correction. The p-value results from DunnTest match the results from pairw.kw. The pairw.kw also provides a confidence interval for the difference in mean ranks between pairs of groups.

The posthoc.kruskal.nemenyi.test function from the **PMCMR** package uses the "Nemenyi" (1963) method of multiple comparisons.

The kruskalmc function from the **pgirmess** package uses the method described by Siegel and Castellan (1988).

It is not clear which method the kruskal function from the **agricolae** package uses. It does not seem to output p-values but it does allow for a wide variety of methods to control the experimentise error rate.

66 Ecoli

Author(s)

Derek H. Ogle, <derek@derekogle.com>, but this is largely a wrapper (see details) for dunn.test in dunn.test written by Alexis Dinno.

References

Dunn, O.J. 1964. Multiple comparisons using rank sums. Technometrics 6:241-252.

See Also

See kruskal.test, dunn.test in **dunn.test**, posthoc.kruskal.nemenyi.test in **PMCMR**, kruskalmc in **pgirmess**, and kruskal in **agricolae**.

Examples

```
## pH in four ponds data from Zar (2010)
ponds <- data.frame(pond=as.factor(rep(1:4,each=8)),</pre>
                     pH=c(7.68,7.69,7.70,7.70,7.72,7.73,7.73,7.76,
                          7.71,7.73,7.74,7.74,7.78,7.78,7.80,7.81,
                          7.74,7.75,7.77,7.78,7.80,7.81,7.84,NA,
                          7.71, 7.71, 7.74, 7.79, 7.81, 7.85, 7.87, 7.91)
ponds <- ponds[complete.cases(ponds),]</pre>
# non-formula usage (default "holm" method)
dunnTest(ponds$pH,ponds$pond)
# formula usage (default "holm" method)
dunnTest(pH~pond,data=ponds)
# other methods
dunnTest(pH~pond,data=ponds,method="bonferroni")
dunnTest(pH~pond, data=ponds, method="bh")
dunnTest(pH~pond, data=ponds, method="none")
# print dunn.test results
tmp <- dunnTest(pH~pond,data=ponds)</pre>
print(tmp,dunn.test.results=TRUE)
```

Ecoli

Population growth of Escherichia coli.

Description

The number of Escherichia coli cells versus time.

expandCounts 67

Format

A data frame with 8 observations on the following 2 variables:

days Elapsed duration of the experiment. **cells** Number of cells in the population.

Source

McKendrick, A.G. and M. Kesava Pai. 1911. The Rate of Multiplication of Micro-Organisms: a Mathematical Study. Proceedings of the Royal Society of Edinburgh. 31: 649-655.

expandCounts

Repeat individual fish data (including lengths) from tallied counts.

Description

Repeat individual fish data, including lengths, from tallied counts and, optionally, add a random digit to length measurements to simulate actual length of fish in the bin. This is useful as a precursor to summaries that require information, e.g., lengths, of individual fish (e.g., length frequency histograms, means lengths).

Usage

```
expandCounts(data, cform, lform = NULL, removeCount = TRUE, lprec = 0.1,
  new.name = "newlen", cwid = 0, verbose = TRUE, ...)
```

Arguments

data	A data.frame that contains variables in cform and 1form.
cform	A formula of the form ~countvar where countvar generically represents the variable in data that contains the counts of individuals. See details.
lform	An optional formula of the form ~lowerbin+upperbin where lowerbin and upperbin generically represent the variables in data that identify the lower-and upper-values of the length bins. See details.
removeCount	A single logical that indicates if the variable that contains the counts of individuals (as given in cform) should be removed form the returned data.frame. The default is TRUE such that the variable will be removed as the returned data.frame contains individuals and the counts of individuals in tallied bins is not relevant to an individual.
lprec	A single numeric that controls the precision to which the random lengths are recorded. See details.
new.name	A single string that contains a name for the new length variable if random lengths are to be created.
cwid	A single positive numeric that will be added to the lower length bin value in instances where the count exceeds one but only a lower (and not an upper) length were recorded. See details.

68 expandCounts

verbose A logical indicating whether progress message should be printed or not.

... Not yet implemented.

Details

Fisheries data may be recorded as tallied counts in the field. For example, field biologists may have simply recorded that there were 10 fish in one group, 15 in another, etc. More specifically, the biologist may have recorded that there were 10 male Bluegill from the first sampling event between 100 and 124 mm, 15 male Bluegill from the first sampling event between 125 and 149 mm, and so on. At times, it may be necessary to expand these counts such that the repeated information appears in individual rows in a new data frame. In this specific example, the tallied counts would be repeated such that the male, Bluegill, first sampling event, 100-124 mm information would be repeated 10 times; the male, Bluegill, first sampling event, 125-149 mm information would be repeated 15 times, and so on. This function facilitates this type of expansion.

Length data has often been collected in a "binned-and-tallied" format (e.g., 10 fish in the 100-124 mm group, 15 in the 125-149 mm group, etc.). This type of data collection does not facilitate easy or precise calculations of summary statistics of length (i.e., mean and standard deviations of length). Expanding the data as described above does not solve this problem because the length data are still essentially categorical (i.e., which group the fish belongs to rather than what it's actual length is). To facilitate computation of summary statistics, the data can be expanded as described above and then a length can be randomly selected from within the recorded length bin to serve as a "measured" length for that fish. This function performs this type of expansion by randomly selecting the length from a uniform distribuion within the length bin (e.g., each value between 100 and 124 mm has the same probability of being selected).

This function makes some assumptions for some coding situtions. First, it assumes that all lowerbin values are actually lower than all upperbin values. The function will throw an error if this is not true. Second, it assumes that if a lowerbin but no upperbin value is given then the lowerbin value is the exact measurement for those fish. Third, it assumes that if an upperbin but no lowerbin value is given that this is a data entry error and that the upperbin value should be the lowerbin value. Fourth, it assumes that it is a data entry error if varcount is zero or NA and lowerbin or upperbin contains values (i.e., why would there be lengths if no fish were captured?).

Value

A data.frame of the same structure as data except that the variable in cform may be deleted and the variable in new.name may be added. The returned data.frame will have more rows than data because of the potential addition of new individuals expanded from the counts in cform.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

See Also

See expandLenFreq for expanding length frequencies where individual fish measurements were made on individual fish in a subsample and the remaining fish were simply counted.

expandCounts 69

```
# all need expansion
( d1 <- data.frame(name=c("Johnson","Johnson","Jones","Frank","Frank","Max"),</pre>
                   lwr.bin=c(15,15.5,16,16,17,17),
                   upr.bin=c(15.5,16,16.5,16.5,17.5,17.5),
                   freq=c(6,4,2,3,1,1)))
expandCounts(d1,~freq)
expandCounts(d1,~freq,~lwr.bin+upr.bin)
# some need expansion
( d2 <- data.frame(name=c("Johnson","Johnson","Jones","Frank","Frank","Max"),</pre>
                   lwr.bin=c(15,15.5,16,16,17.1,17.3),
                   upr.bin=c(15.5,16,16.5,16.5,17.1,17.3),
                   freq=c(6,4,2,3,1,1))
expandCounts(d2,~freq)
expandCounts(d2,~freq,~lwr.bin+upr.bin)
# none need expansion
( d3 <- data.frame(name=c("Johnson","Johnson","Jones","Frank","Frank","Max"),</pre>
                   lwr.bin=c(15,15.5,16,16,17.1,17.3),
                   upr.bin=c(15,15.5,16,16,17.1,17.3),
                   freq=c(6,4,2,3,1,1))
expandCounts(d3,~freq)
expandCounts(d3,~freq,~lwr.bin+upr.bin)
# some need expansion, but different bin widths
( d4 <- data.frame(name=c("Johnson","Johnson","Jones","Frank","Frank","Max"),</pre>
                   lwr.bin=c(15, 15, 16, 16, 17.1,17.3),
                   upr.bin=c(15.5,15.9,16.5,16.9,17.1,17.3),
                   freq=c(6,4,2,3,1,1)) )
expandCounts(d4,~freq)
expandCounts(d4,~freq,~lwr.bin+upr.bin)
# some need expansion but include zeroes and NAs for counts
( d2a <- data.frame(name=c("Johnson","Johnson","Jones","Frank","Frank","Max","Max","Max","Max"),
                    lwr.bin=c(15, 15.5,16 ,16 ,17.1,17.3,NA,NA,NA),
                    upr.bin=c(15.5,16 ,16.5,16.5,17.1,17.3,NA,NA,NA),
                    freq=c(6,4,2,3,1,1,NA,0,NA)))
expandCounts(d2a,~freq,~lwr.bin+upr.bin)
# some need expansion but include NAs for upper values
( d2b <- data.frame(name=c("Johnson","Johnson","Jones","Frank","Frank","Max"),</pre>
                    lwr.bin=c(15, 15.5,16 ,16 ,17.1,17.3),
                    upr.bin=c(NA ,NA ,16.5,16.5,17.1,17.3),
                    freq=c(6,4,2,3,1,1))
expandCounts(d2b,~freq,~lwr.bin+upr.bin)
# some need expansion but include NAs for upper values
( d2c <- data.frame(name=c("Johnson","Johnson","Jones","Frank","Frank","Max"),</pre>
                    lwr.bin=c(NA,NA, 16 ,16 ,17.1,17.3),
                    upr.bin=c(15,15.5,16.5,16.5,17.1,17.3),
                    freq=c(6,4,2,3,1,1))
```

70 extraTests

```
expandCounts(d2c,~freq,~lwr.bin+upr.bin)
## Not run:
\#\#!!\#\#!!\#\# Change path to where example file is and then run to demo
## Read in datafile (note periods in names)
df <- read.csv("c:/aaawork/consulting/R_WiDNR/Statewide/Surveysummaries2010.csv")</pre>
## narrow variables for simplicity
df1 <- df[,c("County","Waterbody.Name","Survey.Year","Gear","Species","Number.of.Fish",
             "Length.or.Lower.Length.IN", "Length.Upper.IN", "Weight.Pounds", "Gender")]
## Sum the count to see how many fish there should be after expansion
sum(df1$Number.of.Fish)
## Simple expansion
df2 <- expandCounts(df1,~Number.of.Fish)</pre>
## Same expansion but include random component to lengths (thus new length variable)
   also note default lprec=0.1
df3 <- expandCounts(df1,~Number.of.Fish,~Length.or.Lower.Length.IN+Length.Upper.IN)
## End(Not run)
```

extraTests

Likelihood ratio and extra sum-of-squares tests.

Description

Likelihood ratio and extra sum-of-squares tests with multiple lm or nls models nested within one common model. This function is most useful when the nested functions are all at the same level; otherise use anova() or lrtest() which are more flexible.

Usage

```
lrt(sim, ..., com, sim.names = sim.name, sim.name = NULL, com.name = NULL)
extraSS(sim, ..., com, sim.names = sim.name, sim.name = NULL,
    com.name = NULL)
## S3 method for class 'extraTest'
print(x, ...)
```

Arguments

sim	The results of one lm or nls model, for example, that is a nested subset of the model in com=.
com	The results of one lm or nls model, for example, that the models in sim= and are a subset of.

extraTests 71

sim.name, sim.names

A string vector of "names" for simple model in sim= and sim.names is preferred but sim.name is allowed to allow for a common typing mistake.

com. name A single "name" string for the complex model in com=.

x An object from lrt() or extraSS().

... More model results that are nested subsets of the model in com=.

Details

anova and lrtest (from lmtest) provide simple methods for conducting extra sum-of-squares or likelihood ratio tests when one model is nested within another model or when there are several layers of simple models all sequentially nested within each other. However, to compare several models that are nested at the same level with one common more complex model, then anova() and lrtest() must be repeated for each comparison. This repetition can be eliminated with lapply() but then the output is voluminous. This function is designed to remove the repetitiveness and to provide output that is compact and easy to read.

Value

The main function returns a matrix with as many rows as model comparisons and columns of the following types:

- Df0 The error degrees-of-freedom from the subset (more simple) model.
- RSSO, logLikO The residual sum-of-squares (from extraSS) or log-likelihood (from lrt) from the subset (more simple) model.
- DfA The error degrees-of-freedom from the com= model.
- RSSA, logLikA The residual sum-of-squares (from extraSS) or log-likelihood (from lrt) from the com= model.
- Df The difference in error degrees-of-freedom between the two models.
- SS, logLik The difference in residual sum-of-squares (from extraSS) or log-likelihood (from lrt) between the two models.
- F, Chisq The corresponding F- (from extraSS) or Chi-square (from 1rt) test statistic.
- Pr(>F), Pr(>Chisq) The corresponding p-value.

Note

This function is experimental at this point. It seems to work fine for lm and nls models. An error will be thrown by extraSS for other model classes, but lrt will not (but it has not been thoroughly tests for other models).

Author(s)

Derek H. Ogle, <derek@derekogle.com>

72 extraTests

```
## Example data
df \leftarrow data.frame(x=c(1,2,3,4,5,6,7,8,9,10),
                  y=c(4,6,5,7,9,8,7,12,16,22),
                  z=as.factor(rep(c("A","B"),each=5)),
                  w=as.factor(rep(c("A","B"),times=5)))
df$x2 <- df$x^2
## Linear (lm()) models
# ... regression
fit.0 <- lm(y^1, data=df)
fit.1 <- lm(y^x, data=df)
fit.2 <- lm(y\sim x2+x, data=df)
extraSS(fit.0,fit.1,com=fit.2)
lrt(fit.0,fit.1,com=fit.2)
# ... show labels for models
extraSS(fit.0,fit.1,com=fit.2,
    sim.names=c("Null Model","Linear"),com.name="Quadratic")
lrt(fit.0,fit.1,com=fit.2,
    sim.names=c("Null Model","Linear"),com.name="Quadratic")
# ... dummy variable regression
fit.2b <- lm(y^x*z,data=df)
extraSS(fit.0,fit.1,com=fit.2b)
lrt(fit.0,fit.1,com=fit.2b)
# ... ANOVAs
fit.1 <- lm(y\sim w, data=df)
fit.2 <- lm(y\sim w \times z, data=df)
extraSS(fit.0,fit.1,com=fit.2)
lrt(fit.0,fit.1,com=fit.2)
## Non-linear (nls()) models
fit.0 = nls(y~c,data=df,start=list(c=10))
fit.1 = nls(y~a*x+c,data=df,start=list(a=1,c=1))
fit.2 = nls(y^b*x^2+a*x+c, data=df, start=list(a=-1, b=0.3, c=10))
extraSS(fit.0,fit.1,com=fit.2)
lrt(fit.0,fit.1,com=fit.2)
## General least-squares (gls()) models
## Not run:
  require(nlme)
  fit.0 <- gls(y~1,data=df,method="ML")</pre>
  fit.1 <- gls(y~x,data=df,method="ML")</pre>
  fit.2 <- gls(y~x2+x,data=df,method="ML")</pre>
  lrt(fit.0,fit.1, com=fit.2)
  ## will return an error ... does not work with gls() models
  # extraSS(fit.0,fit.1, com=fit.2)
## End(Not run)
```

fact2num 73

fact2num

Converts "numeric" factor levels to numeric values.

Description

Converts "numeric" factor levels to numeric values.

Usage

```
fact2num(object)
```

Arguments

object

A vector with "numeric" factor levels to be converted to numeric values.

Value

A numeric vector.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

```
junk <- factor(c(1,7,2,4,3,10))
str(junk)
junk2 <- fact2num(junk)
str(junk2)

## ONLY RUN IN INTERACTIVE MODE
if (interactive()) {

bad <- factor(c("A","B","C"))
# This will result in an error -- levels are not 'numeric'
bad2 <- fact2num(bad)
} ## END IF INTERACTIVE MODE</pre>
```

fishR

Opens web pages associated with the fishR website.

Description

Opens web pages associated with the fishR website in a browser. The user can open the main page or choose a specific page to open.

Usage

```
fishR(where = c("home", "IFAR", "general", "books", "AIFFD", "posts", "news"))
```

Arguments

where

A string that indicates a particular page on the fishR website to open.

Value

None, but a webpage will be opened in the default browswer.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

Examples

```
## ONLY RUN IN INTERACTIVE MODE
if (interactive()) {

fishR()  # home page
fishR("IFAR")  # Introduction to Fisheries Analysis with R page
fishR("general")  # examples page
fishR("books")  # examples page
fishR("AIFFD")  # Analysis & Interpretation of Freshwater Fisheries Data page
fishR("posts")  # blog posts (some examples) page
} ## END IF INTERACTIVE MODE
```

fitPlot

Fitted model plot for an lm, glm, or nls object.

Description

A generic function for constructing a fitted model plot for an lm, glm, or nls object. Supported objects are linear models from simple linear regression (SLR), indicator variable regression (IVR), one-way ANOVA, or two-way ANOVA models; general linear models that are logistic regressions with a binary response; and non-linear regression with a single numerical response variable, at least one continuous explanatory variable and up to two group-factor explanatory variables.

Usage

```
fitPlot(object, ...)
## S3 method for class 'lm'
fitPlot(object, ...)
## S3 method for class 'SLR'
fitPlot(object, plot.pts = TRUE, pch = 16, col.pt = "black",
  col.mdl = "red", lwd = 3, lty = 1, interval = c("none", "confidence",
  "prediction", "both"), conf.level = 0.95, lty.ci = 2, lty.pi = 3,
  xlab = object$Enames[1], ylab = object$Rname, main = "", ...)
## S3 method for class 'IVR'
fitPlot(object, plot.pts = TRUE, pch = c(16, 21, 15, 22, 17,
  24, c(3:14)), col = "rich", lty = c(1:6, 1:6), lwd = 3,
  interval = c("none", "confidence", "prediction", "both"),
  conf.level = 0.95, xlab = object$Enames[1], ylab = object$Rname,
 main = "", legend = "topright", ...)
## S3 method for class 'POLY'
fitPlot(object, ...)
## S3 method for class 'ONEWAY'
fitPlot(object, xlab = object$Enames[1],
 ylab = object$Rname, main = "", type = "b", pch = 16, lty = 1,
  col = "black", interval = TRUE, conf.level = 0.95,
  ci.fun = iCIfp(conf.level), col.ci = col, lty.ci = 1, ...)
## S3 method for class 'TWOWAY'
fitPlot(object, which, change.order = FALSE,
  xlab = object$Enames[ord[1]], ylab = object$Rname, main = "",
  1:6), col = "default", interval = TRUE, conf.level = 0.95,
  ci.fun = iCIfp(conf.level), lty.ci = 1, legend = "topright",
  cex.leg = 1, box.lty.leg = 0, ...)
## S3 method for class 'nls'
fitPlot(object, d, pch = c(19, 1), col.pt = c("black", "red"),
  col.mdl = col.pt, lwd = 2, lty = 1, plot.pts = TRUE,
  jittered = FALSE, ylim = NULL, legend = FALSE,
  legend.lbls = c("Group 1", "Group 2"), ylab = names(mdl$model)[1],
  xlab = names(mdl$model)[xpos], main = "", ...)
## S3 method for class 'glm'
fitPlot(object, ...)
## S3 method for class 'logreg'
fitPlot(object, xlab = names(object$model)[2],
```

```
ylab = names(object$model)[1], main = "", plot.pts = TRUE,
col.pt = "black", transparency = NULL, plot.p = TRUE, breaks = 25,
p.col = "blue", p.pch = 3, p.cex = 1, yaxis1.ticks = seq(0, 1, 0.1),
yaxis1.lbls = c(0, 0.5, 1), yaxis2.show = TRUE, col.mdl = "red",
lwd = 2, lty = 1, mdl.vals = 50, xlim = range(x), ...)
```

Arguments

guments	
object	An lm or nls object (i.e., returned from fitting a model with either lm or nls).
plot.pts	A logical that indicates (TRUE (default)) whether the points are plotted along with the fitted lines. Set to FALSE to plot just the fitted lines.
pch	A numeric or vector of numerics that indicates what plotting characther codes should be used. In SLR this is the single value to be used for all points. In IVR a vector is used to identify the characters for the levels of the second factor.
col.pt	A string used to indicate the color of the plotted points. Used only for SLR and logistic regression objects.
col.mdl	A string used to indicate the color of the fitted line. Used only for SLR and logistic regression objects.
lwd	A numeric used to indicate the line width of the fitted line.
lty	A numeric or vector of numerics used to indicate the type of line used for the fitted line. In SLR this is a single value to be used for the fitted line. In IVR a vector is used to identify the line types for the levels of the second factor. See par.
interval	In SLR or IVR, a string that indicates whether to plot confidence (="confidence") or prediction (="prediction") intervals. For a SLR object both can be plotted by using ="both". In one-way or two-way ANOVA, a logical that indicates whether the confidence intervals should be plotted or not.
conf.level	A decimal numeric that indicates the level of confidence to use for confidence and prediction intervals.
lty.ci	a numeric used to indicate the type of line used for the confidence band lines for SLR objects or interval lines for one-way and two-way ANOVA. For IVR, the confidence band types are controlled by 1ty.
lty.pi	a numeric used to indicate the type of line used for the prediction band lines for SLR objects. For IVR, the prediction band types are controlled by 1ty. See par.
xlab	a string for labelling the x-axis.
ylab	a string for labelling the y-axis.
main	a string for the main label to the plot. Defaults to the model call.
col	A vector of color names or numbers or the name of a paletted (see details) that indicates what color of points and lines to use for the levels of the first factor in an IVR or the second factor in a two-way ANOVA.
legend	Controls use and placement of the legend. See details.
type	The type of graphic to construct in a one-way and two-way ANOVA. If "b" then points are plotted and lines are used to connect points (DEFAULT). If "p" then only points are used and if "1" then only lines are drawn.

ci.fun	A function used to put error bars on the one-way or two-way ANOVA graphs. The default is to use the internal iCIfp function which will place t-distribution based confidence intervals on the graph. The user can provide alternative functions that may plot other types of 'error bars'. See examples in lineplot.CI function of sciplot package.
col.ci	A vector of color names or numbers or the name of a palette (see details) that indicates what colors to use for the confidence interval bars in one-way and two-way ANOVAs.
which	A character string listing the factor in the two-way ANOVA for which the means should be calculated and plotted. This argument is used to indicate for which factor a main effects plot should be constructed. If left missing then an interaction plot is constructed.
change.order	A logical that is used to change the order of the factors in the 1m object. This is used to change which factor is plotted on the x-axis and which is used to connect the means when constructing an interaction plot (ignored if which is used).
cex.leg	A single numeric values used to represent the character expansion value for the legend. Ignored if legend=FALSE.
box.lty.leg	A single numeric values used to indicate the type of line to use for the box around the legend. The default is to not plot a box.
d	A data frame that contains the variabls used in construction of the nls object.
jittered	A logical that indicates whether the points should be jittered horizontally.
ylim	A vector of length two to control the y-axis in the nonlinear regression plot.
legend.lbls	A vector of strings that will be the labels for the legend in an nls fitPlot graphic.
transparency	A numeric that indicates how many points would be plotted on top of each other in a logistic regression before the 'point' would have the full pt.col color. The reciprocal of this value is the alpha transparency value.
plot.p	A logical that indicates if the proportion for categorized values of \boldsymbol{X} are plotted (TRUE; default).
breaks	A number that indicates how many intervals over which to compute proportions or a numeric vector that contains the endpoints of the intervals over which to compute proportions if plot.p=TRUE.
p.col	A color to plot the proportions.
p.pch	A plotting character for plotting the proportions.
p.cex	A character expansion factor for plotting the proportions.
yaxis1.ticks	A numeric vector that indicates where tick marks should be placed on the left y-axis (for the proportion of 'successes') for the logistic regression plot.
yaxis1.lbls	A numeric vector that indicates labels for the tick marks on the left y-axis (for the proportion of 'successes') for the logistic regression plot.
yaxis2.show	A logical that indicates whether the right y-axis should be created (=TRUE; default) or not for the logistic regression plot.
mdl.vals	A numer representing the number of values to use for plotting the logistic regression. A larger number means a smoother line.

xlim A vector of length two to control the x-axis in the logistic regression plot. If this

is changed from the default then the domain over which the logistic regression

model is plotted will change.

... Other arguments to be passed to the plot functions.

Details

This function does not work with a multiple linear regression, indicator variable regressions with more than two factors, ANOVAs other than one-way and two-way, or models with a categorical response variable. In addition, if the linear model contains a factor then the model must be fit with the quantitative explanatory variable first, followed by the factor(s). This function only works for non-linear models with two or fewer groups.

This function is basically a wrapper to a variety of other functions. For one-way or two-way ANOVAs the primary functions called are interaction.plot and lineplot.CI. For simple linear regression the function performs similarly to abline except that the line is constrained to the domain. For indicator variable regression the function behaves as if several abline functions had been called.

A legend can be added to the plot in three different ways. First, if legend = TRUE then the R console is suspended until the user places the legend on the graphic by clicking on the graphic at the point where the upper-left corner of the legend should appear. Second, the legend= argument can be set to one of "bottomright", "bottom", "bottomleft", "left", "topleft", "top", "topright", "right" and "center". In this case, the legend will be placed inside the plot frame at the given location. Finally, the legend= argument can be set to a vector of length two which identifies the plot coordinates for the upper-left corner of where the legend should be placed. A legend will not be drawn if legend = FALSE or legend = NULL. A legend also will not be drawn if there are not multiple groups in the model.

Value

None. However, a fitted-line plot is produced.

Note

This function is meant to allow newbie students the ability to visualize the most common linear models found in an introductory or intermediate level undergraduate statistics course without getting "bogged-down" in the gritty details of a wide variety of functions. This generic function and it's S3 functions allow the student to visualize the means plot of a one-way anova, the main effects and interaction plots of a two-way ANOVA, the fit of a simple linear regression, the fits of many lines in an indicator variable regression, and the fit of a non-linear model with a simple and mostly common set of arguments – generally, all that is required is a fitted linear model of the type mentioned here as the first argument. This function thus allows newbie students to interact with and visualize moderately complex linear models in a fairly easy and efficient manner. THIS IS NOT A RESEARCH GRADE FUNCTION and the user should learn how to use the functions that this function is based on, build plots from "scratch", or use more sophisticated plotting packages (e.g., ggplot2 or lattice).

Author(s)

Derek H. Ogle, <derek@derekogle.com>

See Also

See abline, regLine in **car**, plotmeans in **gplots**, error.bars in **psych**, interaction.plot, and lineplot.CI in **sciplot** for similar functionality. See residPlot for related functionality.

```
## Linear models example
data(Mirex)
Mirex$year <- factor(Mirex$year)</pre>
# Indicator variable regression with two factors
lm1 <- lm(mirex~weight*year*species,data=Mirex)</pre>
fitPlot(lm1)
fitPlot(lm1,ylim=c(0,0.65),legend="topleft")
# Indicator variable regression with one factor (also showing confidence bands)
lm2 <- lm(mirex~weight*year,data=Mirex)</pre>
fitPlot(lm2,legend="topleft")
fitPlot(lm2,legend="topleft",interval="c")
# Indicator variable regression with one factor (assuming parallel lines)
lm3 <- lm(mirex~weight+year,data=Mirex)</pre>
fitPlot(lm3,legend="topleft")
fitPlot(lm3,legend="topleft",pch=5)
# Simple linear regression (showing color change and confidence and prediction bands)
lm4 <- lm(mirex~weight,data=Mirex)</pre>
fitPlot(lm4,col.mdl="blue")
fitPlot(lm4,interval="both")
# One-way ANOVA
lm5 <- lm(mirex~year,data=Mirex)</pre>
fitPlot(lm5)
# Two-way ANOVA (showing both interaction plots and a color change)
lm6 <- lm(mirex~year*species,data=Mirex)</pre>
fitPlot(lm6,legend="bottomleft")
fitPlot(lm6,change.order=TRUE)
fitPlot(lm6,change.order=TRUE,col="jet")
# Two-way ANOVA (showing main effects plots)
fitPlot(lm6, which="species")
fitPlot(lm6,which="year")
# Polynomial regression
lm7 <- lm(mirex~weight+I(weight^2),data=Mirex)</pre>
fitPlot(lm7,interval="both")
## Non-linear model example
data(Ecoli)
lr.sv <- list(B1=6,B2=7.2,B3=-1.5)</pre>
nl1 <- nls(cells~B1/(1+exp(B2+B3*days)),start=lr.sv,data=Ecoli)</pre>
```

80 FSA

```
fitPlot(nl1,Ecoli,cex.main=0.7,lwd=2)
fitPlot(nl1,Ecoli,xlab="Day",ylab="Cellsx10^6/ml",plot.pts=FALSE)

## Logistic regression example
## NASA space shuttle o-ring failures -- from graphics package
fail <- factor(c(2,2,2,2,1,1,1,1,1,2,1,2,1,1,1,1,1,1),
levels = 1:2, labels = c("no", "yes"))
temperature <- c(53,57,58,63,66,67,67,67,68,69,70,70,70,70,72,73,75,75,76,76,78,79,81)
d <- data.frame(fail,temperature)
glm1 <- glm(fail~temperature,data=d,family="binomial")
fitPlot(glm1)
fitPlot(glm1, yaxis1.ticks=c(0,1),yaxis1.lbls=c(0,1))
# changing the size of the y-axis labels
par(cex.axis=1.5,cex.lab=1.5)
fitPlot(glm1)</pre>
```

FSA

Fisheries stock assessment methods and data.

Description

Functions to support basic fisheries stock assessment methods.

Details

Functions from this package can be used to perform a variety of basic fisheries stock assessment methods. Detailed descriptions for most functions are available in the Introductory to Fisheries Analysis with R book (Ogle 2016). Vignettes for the boxed examples in the "Analysis and Interpretation of Freshwater Fisheries Data" book can be viewed with fishR("AIFFD").

Questions, comments, or suggestions should be given on the GitHub FSA Issues page.

Packages with related functionality by the same author are

- The FSAdata package contains additional data sets.
- The FSAsim package simulation routines for various fisheries methods.
- The FSAWs package contains functions for developing and validating standard weight equations.

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

fsaNews 81

fsaNews

Read news and changes for the 'FSA' package.

Description

Opens up the News.md GitHub file for the 'FSA' package in an external browser.

Usage

```
fsaNews()
FSANews()
```

Value

None.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

Examples

```
## ONLY RUN IN INTERACTIVE MODE
if (interactive()) {
fsaNews()
FSANews()
} ## END IF INTERACTIVE MODE
```

FSAUtils

Capitalizes the first letter of first or all words in a string.

Description

Capitalizes the first letter of first or all words in a string.

Usage

```
capFirst(x, which = c("all", "first"))
```

Arguments

x A single string.

which A single string that indicates whether all (the default) or only the first words

should be capitalized.

82 gConvert

Value

A single string with the first letter of the first or all words capitalized.

Author(s)

```
Derek H. Ogle, <derek@derekogle.com>
```

Examples

```
## Capitalize first letter of all words (the default)
capFirst("Derek Ogle")
capFirst("derek ogle")
capFirst("derek")
## Capitalize first letter of only the first words
capFirst("Derek Ogle",which="first")
capFirst("derek ogle", which="first")
capFirst("derek", which="first")
## apply to all elements in a vector
vec <- c("Derek Ogle","derek ogle","Derek ogle","derek Ogle","DEREK OGLE")</pre>
capFirst(vec)
capFirst(vec,which="first")
## check class types
class(vec)
vec1 <- capFirst(vec)</pre>
class(vec1)
fvec <- factor(vec)</pre>
fvec1 <- capFirst(fvec)</pre>
class(fvec1)
```

gConvert

Converts between types of measurements from calcified structures.

Description

Converts wide format growth data from radial to incremental or incremental to radial measurements.

Usage

```
gConvert(df, in.pre = NULL, in.var = NULL, out.type = c("inc", "rad"),
  out.pre = out.type)
```

Arguments

df A data.frame that contains the growth measurement data in wide or one-fish-

per-line format.

in.pre A string that indicates the prefix for all variable names in the input data frame

that contain the measurements from the calcified structures. See details.

gConvert 83

in.var	A vector of column numbers or variable names in the input data frame that contain the measurements from the calcified structures. See details.
out.type	A string that identifies the output format data type (i.e., the format to convert to). If "inc} (the default) the output data frame will be incremental measurements. If \co the output data frame will be radial measurements.
out.pre	A string that indicates the prefix to use for the newly computed measurements in the output data frame. Defaults to the same string as out, type

Details

The data must be in WIDE format where each row contains all information (including all measurements from the calcified structure) for an individual fish. It is assumed that the input data frame is of the opposite data type given in out. type (i.e., that a conversion is needed). It does not check to see if this is true.

The columns that contain the original measurement data can specified in a variety of ways. First, if all columns begin with the same prefix (and no other columns contain that prefix), then the prefix string may be given to in.pre=. Second, a sequence of column numbers may be given to in.var= with the #:# (if the columns are contiguous) or as a vector (if the columns are not contiguous). Third, a vector of column names may be given to in.var=. Note that one, but not both, of in.var= or in.pre= must be specified by the user.

The newly computed data will be labeled with a prefix the same as out.type=(i.e., "rad" or "inc") unless out.pre= is set by the user. For example, if the data are converted to radial measurements, then the output variables will be "rad1", "rad2", etc. unless out.pre= was changed from the default. This function assumes that the measurements start with age-1.

Value

A data frame with all columns, except for those defined by in.pre or in.var, from the df retained as the left-most columns and the original data in the in.var columns converted to the out.type type as the remaining columns.

Author(s)

```
Derek H. Ogle, <derek@derekogle.com>
```

See Also

See addRadCap for related functionality.

```
## Get data with radial measurements
data(SMBassWB)
headtail(SMBassWB)
## Use in.pre= to convert to increments
SMBi1 <- gConvert(SMBassWB,in.pre="anu",out.type="inc")</pre>
headtail(SMBi1)
```

growthModels

Creates a function for a specific parameterization of the von Bertalanffy, Gompertz, Richards, and logistic growth functions.

Description

Creates a function for a specific parameterizations of the von Bertalanffy, Gompertz, Richards, and logistic growth functions. Use vbModels(), GompertzModels(), RichardsModels(), and logisticModels() to see the equations for each growth function.

Usage

```
vbFuns(type = c("Typical", "typical", "BevertonHolt", "Original", "original",
    "vonBertalanffy", "GQ", "GallucciQuinn", "Mooij", "Weisberg", "Schnute",
    "Francis", "Laslett", "Polacheck", "Somers", "Somers2", "Fabens2",
    "Wang", "Wang2", "Wang3"), simple = FALSE, msg = FALSE)

GompertzFuns(type = c("Ricker1", "Ricker2", "Ricker3", "QD1", "QD2", "QD3",
    "KM", "AFS", "original", "Troynikov1", "Troynikov2"), simple = FALSE,
    msg = FALSE)

RichardsFuns(type = 1, simple = FALSE, msg = FALSE)

logisticFuns(type = c("CJ1", "CJ2", "Karkach", "HaddonI"), simple = FALSE,
    msg = FALSE)

vbModels(family = c("size", "seasonal", "tagging"), cex = 1, ...)

GompertzModels(family = c("size", "tagging"), cex = 1.25, ...)

RichardsModels(cex = 1, ...)

logisticModels(family = c("size", "tagging"), cex = 1.25, ...)
```

Arguments

type	A string (for von Bertalanffy, Gompertz, and logistic) or numeric (for Richards) that indicates the specific parameterization of the growth function.
simple	A logical that indicates whether the function will accept all parameter values in the first parameter argument (=FALSE; DEFAULT) or whether all individual parameters must be specified in separate arguments (=TRUE).
msg	A logical that indicates whether a message about the growth function and parameter definitions should be output (=TRUE) or not (=FALSE; DEFAULT).
family	A string that indicates the family of growth functions to display. Choices are "size" (DEFAULT), "tagging", and "seasonal" (only for the von Bertalanffy models).
cex	A single numeric character expansion value.
	Not implemented.

Value

The functions ending in Funs return a function that can be used to predict fish size given a vector of ages and values for the growth function parameters and, in some parameterizations, values for some constants. The result should be saved to an object that is then the function name. When the resulting function is used, the parameters are ordered as shown when the definitions of the parameters are printed after the function is called (if msg=TRUE). If simple=FALSE, then the values for all parameters may be included as a vector in the first parameter argument. Similarly, the values for all constants may be included as a vector in the first constant argument (i.e., t1). If simple=TRUE, then all parameters and constants must be declared individually. The resulting function is somewhat easier to read when simple=TRUE, but is less general for some applications.

The functions ending in Models return a graphic that uses plotmath to show the growth function equations in a pretty format.

IFAR Chapter

12-Individual Growth.

Note

Take note of the folling for each type of growth function:

- · von Bertalanffy
 - The 'original' and 'vonBertalanffy' and the 'typical' and 'BevertonHolt' parameterizations are synonymous.
- Gompertz
 - The 'Ricker2' and 'QD1'; 'Ricker3' and 'QD2'; and 'Ricker1', 'AFS', and 'KM' parameterizations are synonymous in their usage here.
 - The parameterizations and parameters for the Gompertz function are varied and confusing in the literature. I have attempted to use a uniform set of paraemters in these functions but that makes a direct comparison to the literature difficult. Common sources for Gompertz models are listed in the references below. I make some comments here to aid the

- comparison. It is likely worth your while to look at GompertzModels while you make these comparisons.
- Within FSA, L0 is the mean length at age 0, Linf is the mean asymptotic length, ti is the age at the inflection point, gi is the instantaneous growth rate at the inflection point, t* is a dimensionless parameter related to time/age, and a is a dimensionless parameter related to growth.
- In the Quinn and Deriso (1999) models (the 'QD' models), the a parameter here is equal to lambda/K there and the gi parameter here is equal to the K parameter there. Also note that their Y is L here.
- In the Ricker (1979)[p. 705] models (the 'Ricker' models), the a parameter here is equal to k there and the gi parameter here is equal to the g parameter there. Also note that their w is L here. In the Ricker (1979) models as presented in Campana and Jones (1992), the a parameter here is equal to k there and the gi parameter here is equal to the G parameter there. Also note that their X is L here.
- The model in Ricker (1975)[p. 232] is the same as 'Ricker2' where the a parameter here is qual to G there and the gi parameter here is equal to the g parameter there. Also note that their w is L here.
- The model in Quist et al. (2012)[p. 714] is the same as 'Ricker1' where the gi parameter here is equal to G there and the ti parameter here is equal to the t0 parameter there. This parameterization can also be called with type="AFS".
- The model in Katsanevakis and Maravelias (2008) is the same as 'Ricker1' where the gi parameter here is equal to k2 there and the ti parameter here is equal to t2 there. This parameterization can also be called with type="KM".

· Richards

- Within FSA, Linf is the mean asymptotic length, ti is the age at the inflection point, k is related to growth (slope at the inflection point), b is related to the vertical position of the inflection point, and L0 is the mean length at age-0.
- The parameterizations (1-6) correspond to models/equations 1, 4, 5, 6, 7, and 8, respectively, in Tjorve and Tjorve (2010). Note that their A, S, k, d, and B are Linf, a, k, b, and L0 here (in FSA).

logistic

- Within FSA, L0 is the mean length at age 0, Linf is the mean asymptotic length, ti is the age at the inflection point, and gninf is the instantaneous growth rate at negative infinity.

Author(s)

Derek H. Ogle, <derek@derekogle.com>, thanks to Gabor Grothendieck for a hint about using get().

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL. Campana, S.E. and C.M. Jones. 1992. Analysis of otolith microstructure data. Pages 73-100 In D.K. Stevenson and S.E. Campana, editors. Otolith microstructure examination and analysis. Canadian Special Publication of Fisheries and Aquatic Sciences 117.

Fabens, A. 1965. Properties and fitting of the von Bertalanffy growth curve. Growth 29:265-289.

Francis, R.I.C.C. 1988. Are growth parameters estimated from tagging and age-length data comparable? Canadian Journal of Fisheries and Aquatic Sciences, 45:936-942.

Gallucci, V.F. and T.J. Quinn II. 1979. Reparameterizing, fitting, and testing a simple growth model. Transactions of the American Fisheries Society, 108:14-25.

Garcia-Berthou, E., G. Carmona-Catot, R. Merciai, and D.H. Ogle. A technical note on seasonal growth models. Reviews in Fish Biology and Fisheries 22:635-640.

Gompertz, B. 1825. On the nature of the function expressive of the law of human mortality, and on a new method of determining the value of life contingencies. Philosophical Transactions of the Royal Society of London. 115:513-583.

Haddon, M., C. Mundy, and D. Tarbath. 2008. Using an inverse-logistic model to describe growth increments of blacklip abalone (*Haliotis rubra*) in Tasmania. Fishery Bulletin 106:58-71.

Karkach, A. S. 2006. Trajectories and models of individual growth. Demographic Research 15:347-400.

Katsanevakis, S. and C.D. Maravelias. 2008. Modelling fish growth: multi-model inference as a better alternative to a priori using von Bertalanffy equation. Fish and Fisheries 9:178-187.

Mooij, W.M., J.M. Van Rooij, and S. Wijnhoven. 1999. Analysis and comparison of fish growth from small samples of length-at-age data: Detection of sexual dimorphism in Eurasian perch as an example. Transactions of the American Fisheries Society 128:483-490.

Polacheck, T., J.P. Eveson, and G.M. Laslett. 2004. Increase in growth rates of southern bluefin tuna (*Thunnus maccoyii*) over four decades: 1960 to 2000. Canadian Journal of Fisheries and Aquatic Sciences, 61:307-322.

Quinn, T. J. and R. B. Deriso. 1999. Quantitative Fish Dynamics. Oxford University Press, New York, New York. 542 pages.

Quist, M.C., M.A. Pegg, and D.R. DeVries. 2012. Age and Growth. Chapter 15 in A.V. Zale, D.L Parrish, and T.M. Sutton, Editors Fisheries Techniques, Third Edition. American Fisheries Society, Bethesda, MD.

Richards, F. J. 1959. A flexible growth function for empirical use. Journal of Experimental Biology 10:290-300.

Ricker, W.E. 1975. Computation and interpretation of biological statistics of fish populations. Technical Report Bulletin 191, Bulletin of the Fisheries Research Board of Canada.

Ricker, W.E. 1979. Growth rates and models. Pages 677-743 In W.S. Hoar, D.J. Randall, and J.R. Brett, editors. Fish Physiology, Vol. 8: Bioenergetics and Growth. Academic Press, NY, NY.

Schnute, J. 1981. A versatile growth model with statistically stable parameters. Canadian Journal of Fisheries and Aquatic Sciences, 38:1128-1140.

Somers, I. F. 1988. On a seasonally oscillating growth function. Fishbyte 6(1):8-11.

Tjorve, E. and K. M. C. Tjorve. 2010. A unified approach to the Richards-model family for use in growth analyses: Why we need only two model forms. Journal of Theoretical Biology 267:417-425.

Troynikov, V. S., R. W. Day, and A. M. Leorke. Estimation of seasonal growth parameters using a stochastic Gompertz model for tagging data. Journal of Shellfish Research 17:833-838.

Vaughan, D. S. and T. E. Helser. 1990. Status of the red drum stock of the Atlantic coast: Stock assessment report for 1989. NOAA Technical Memorandum NMFS-SEFC-263, 117 p.

Wang, Y.-G. 1998. An improved Fabens method for estimation of growth parameters in the von Bertalanffy model with individual asymptotes. Canadian Journal of Fisheries and Aquatic Sciences 55:397-400.

Weisberg, S., G.R. Spangler, and L. S. Richmond. 2010. Mixed effects models for fish growth. Canadian Journal of Fisheries And Aquatic Sciences 67:269-277.

Winsor, C.P. 1932. The Gompertz curve as a growth curve. Proceedings of the National Academy of Sciences. 18:1-8.

See Also

See Schnute for an implementation of the Schnute (1981) model.

```
## See the equations
## Not run: windows(8,5)
vbModels()
## Not run: windows(6,5)
vbModels("seasonal")
vbModels("tagging")
## Not run: windows(5,5)
GompertzModels()
GompertzModels("tagging")
RichardsModels()
logisticModels()
## Simple Examples -- Von B
( vb1 <- vbFuns() )</pre>
ages <- 0:20
plot(vb1(ages,Linf=20,K=0.3,t0=-0.2)~ages,type="b",pch=19)
( vb2 <- vbFuns("Francis") )</pre>
plot(vb2(ages,L1=10,L2=19,L3=20,t1=2,t3=18)~ages,type="b",pch=19)
( vb2c <- vbFuns("Francis", simple=TRUE) ) # compare to vb2
## Simple Examples -- Gompertz
( gomp1 <- GompertzFuns() )</pre>
plot(gomp1(ages,Linf=800,gi=0.5,ti=5)~ages,type="b",pch=19)
( gomp2 <- GompertzFuns("Ricker2") )</pre>
plot(gomp2(ages, L0=2, a=6, gi=0.5)^ages, type="b", pch=19)
( gomp2c <- GompertzFuns("Ricker2", simple=TRUE) ) # compare to gomp2</pre>
( gompT <- GompertzFuns("Troynikov1"))</pre>
## Simple Examples -- Richards
( rich1 <- RichardsFuns() )</pre>
plot(rich1(ages,Linf=800,k=0.5,a=1,b=6)~ages,type="b",pch=19)
( rich2 <- RichardsFuns(2) )</pre>
plot(rich2(ages, Linf=800, k=0.5, ti=3, b=6)^ages, type="b", pch=19)
( rich3 <- RichardsFuns(3) )</pre>
plot(rich3(ages,Linf=800,k=0.5,ti=3,b=0.15)~ages,type="b",pch=19)
```

```
( rich4 <- RichardsFuns(4) )</pre>
plot(rich4(ages,Linf=800,k=0.5,ti=3,b=0.95)~ages,type="b",pch=19)
lines(rich4(ages,Linf=800,k=0.5,ti=3,b=1.5)~ages,type="b",pch=19,col="blue")
( rich5 <- RichardsFuns(5) )</pre>
plot(rich5(ages,Linf=800,k=0.5,L0=50,b=1.5)~ages,type="b",pch=19)
( rich6 <- RichardsFuns(6) )</pre>
plot(rich6(ages,Linf=800,k=0.5,ti=3,Lninf=50,b=1.5)~ages,type="b",pch=19)
( rich2c <- RichardsFuns(2,simple=TRUE) ) # compare to rich2</pre>
## Simple Examples -- Logistic
( log1 <- logisticFuns() )</pre>
plot(log1(ages,Linf=800,gninf=0.5,ti=5)~ages,type="b",pch=19)
( log2 <- logisticFuns("CJ2") )
plot(log2(ages,Linf=800,gninf=0.5,a=10)~ages,type="b",pch=19)
( log2c \leftarrow logisticFuns("CJ2", simple=TRUE) ) # compare to log2
( log3 <- logisticFuns("Karkach") )</pre>
plot(log3(ages,L0=10,Linf=800,gninf=0.5)~ages,type="b",pch=19)
( log4 <- logisticFuns("HaddonI") )</pre>
## Examples of fitting
## After the last example a plot is constructed with three
## or four lines on top of each other illustrating that the
    parameterizations all produce the same fitted values.
##
    However, observe the correlations in the summary() results.
## Von B
data(SpotVA1)
# Fitting the typical paramaterization of the von B function
fit1 <- nls(tl~vb1(age,Linf,K,t0),data=SpotVA1,start=vbStarts(tl~age,data=SpotVA1))</pre>
summary(fit1,correlation=TRUE)
plot(tl~age,data=SpotVA1,pch=19)
curve(vb1(x,Linf=coef(fit1)),from=0,to=5,col="red",lwd=10,add=TRUE)
# Fitting the Francis paramaterization of the von B function
fit2 <- nls(tl\sim b2c(age,L1,L2,L3,t1=0,t3=5),data=SpotVA1,
            start=vbStarts(tl~age,data=SpotVA1,type="Francis",ages2use=c(0,5)))
summary(fit2,correlation=TRUE)
curve(vb2c(x,L1=coef(fit2)[1],L2=coef(fit2)[2],L3=coef(fit2)[3],t1=0,t3=5),
      from=0, to=5, col="blue", lwd=5, add=TRUE)
# Fitting the Schnute parameterization of the von B function
vb3 <- vbFuns("Schnute")</pre>
fit3 <- nls(tl\sim vb3(age,L1,L3,K,t1=0,t3=4),data=SpotVA1,
            start=vbStarts(tl^age, data=SpotVA1, type="Schnute", ages2use=c(0,4)))
summary(fit3,correlation=TRUE)
curve(vb3(x,L1=coef(fit3),t1=c(0,4)),from=0,to=5,col="green",lwd=2,add=TRUE)
## Gompertz
# Make some fake data using the original parameterization
gomp0 <- GompertzFuns("original")</pre>
# setup ages, sample sizes (general reduction in numbers with
```

```
# increasing age), and additive SD to model
n < c(10,40,35,25,12,10,10,8,6,5,3,3,3,2,2)
sd <- 15
# expand ages
ages <- rep(t,n)
# get lengths from gompertz and a random error for individuals
lens <- gompO(ages,Linf=450,a=1,gi=0.3)+rnorm(length(ages),0,sd)</pre>
# put together as a data.frame
df <- data.frame(age=ages,len=round(lens,0))</pre>
# Fit first Ricker parameterization
fit1 <- nls(len~gomp1(age,Linf,gi,ti),data=df,start=list(Linf=500,gi=0.3,ti=3))</pre>
summary(fit1,correlation=TRUE)
plot(len~age, data=df, pch=19, col=rgb(0,0,0,1/5))
curve(gomp1(x,Linf=coef(fit1)),from=0,to=15,col="red",lwd=10,add=TRUE)
# Fit third Ricker parameterization
fit2 <- nls(len~gomp2(age,L0,a,gi),data=df,start=list(L0=30,a=3,gi=0.3))</pre>
summary(fit2,correlation=TRUE)
curve(gomp2(x,L0=coef(fit2)),from=0,to=15,col="blue",lwd=5,add=TRUE)
# Fit third Quinn and Deriso parameterization (using simple=TRUE model)
gomp3 <- GompertzFuns("QD3",simple=TRUE)</pre>
fit3 <- nls(len~gomp3(age,Linf,gi,t0),data=df,start=list(Linf=500,gi=0.3,t0=0))</pre>
summary(fit3,correlation=TRUE)
curve(gomp3(x,Linf=coef(fit3)[1],gi=coef(fit3)[2],t0=coef(fit3)[3]),
      from=0, to=15, col="green", lwd=2, add=TRUE)
## Richards
# Fit first Richards parameterization
fit1 <- nls(len~rich1(age,Linf,k,a,b),data=df,start=list(Linf=450,k=0.25,a=0.65,b=3))</pre>
summary(fit1,correlation=TRUE)
plot(len~age, data=df, pch=19, col=rgb(0,0,0,1/5))
curve(rich1(x,Linf=coef(fit1)),from=0,to=15,col="red",lwd=10,add=TRUE)
# Fit second Richards parameterization
fit2 <- nls(len~rich2(age,Linf,k,ti,b),data=df,start=list(Linf=450,k=0.25,ti=3,b=3))
summary(fit2,correlation=TRUE)
curve(rich2(x,Linf=coef(fit2)),from=0,to=15,col="blue",lwd=7,add=TRUE)
# Fit third Richards parameterization
fit3 <- nls(len rich3(age, Linf, k, ti, b), data = df, start = list(Linf = 450, k = 0.25, ti = 3, b = -0.3))
summary(fit3,correlation=TRUE)
curve(rich3(x,Linf=coef(fit3)),from=0,to=15,col="green",lwd=4,add=TRUE)
# Fit fourth Richards parameterization
fit4 <- nls(len~rich4(age,Linf,k,ti,b),data=df,start=list(Linf=450,k=0.25,ti=3,b=0.7))
summary(fit4,correlation=TRUE)
curve(rich4(x,Linf=coef(fit4)),from=0,to=15,col="black",lwd=1,add=TRUE)
## Logistic
# Fit first Campana-Jones parameterization
```

headtail 91

headtail

Shows rows from the head and tail of a data frame or matrix.

Description

Shows rows from the head and tail of a data frame or matrix.

Usage

```
headtail(x, n = 3L, which = NULL, addrownums = TRUE, ...)
```

Arguments

x	A data frame or matrix.
n	A single numeric that indicates the number of rows to display from each of the head and tail of structure.
which	A numeric or string vector that contains the column numbers or names to display. Defaults to showing all columns.
addrownums	If there are no row names for the MATRIX, then create them from the row numbers.
	Arguments to be passed to or from other methods.

Value

A matrix or data.frame with 2*n rows.

Note

If n is larger than the number of rows in x then all of x is displayed.

92 hist.formula

Author(s)

Derek H. Ogle, <derek@derekogle.com>

Examples

```
headtail(iris)
headtail(iris,10)
headtail(iris,which=c("Sepal.Length","Sepal.Width","Species"))
headtail(iris,which=grep("Sepal",names(iris)))
headtail(iris,n=200)
## Make a matrix for demonstration purposes only
miris <- as.matrix(iris[,1:4])</pre>
headtail(miris)
headtail(miris, 10)
headtail(miris,addrownums=FALSE)
headtail(miris, 10, which=2:4)
## Make a tbl_df type from dplyr ... note how headtail()
## is not limited by the tbl_df restriction on number of
## rows to show (but head() is).
if (require(dplyr)) {
  iris2 <- tbl_df(iris)</pre>
  class(iris2)
  headtail(iris2, n=15)
  head(iris2, n=30)
}
```

hist.formula

Creates separate histograms by levels.

Description

Creates separate histograms of a quantitative variable by levels of a factor variable.

Usage

```
## S3 method for class 'formula'
hist(formula, data = NULL, main = "", right = FALSE,
    pre.main = "", xlab = NULL, ylab = "Frequency", same.breaks = TRUE,
    same.ylim = TRUE, ymax = NULL, col = "gray90",
    nrow = round(sqrt(num)), ncol = ceiling(sqrt(num)), byrow = TRUE,
    iaxs = TRUE, ...)
```

Arguments

formula A formula. See details.

data An optional data frame that contains the variables in the model.

hist.formula 93

main	A character string used as the main title for when a SINGLE histogram is produced.
right	A logical that indicates if the histogram bins are right-closed (left open) intervals (=TRUE) or not (=FALSE; default).
pre.main	A character string to be used as a prefix for the main title when multiplie histograms are produced. See details.
xlab	A character label for the x-axis. Defaults to name of quantitative variable in formula.
ylab	A character label for the y-axis. Defaults to "Frequency".
same.breaks	A logical that indicates whether the same break values (i.e., bins) should be used on each histogram. Defaults to TRUE.
same.ylim	A logicial that indicates whether the same limits for the y-axis should be used on each histogram. Defaults to TRUE.
ymax	A single value that sets the maximum y-axis limit for each histogram or a vector of length equal to the number of groups that sets the maximum y-axis limit for each histogram separately. If NULL (default), then a value will be found.
col	A string that indicates the color for the bars on the histogram. Defaults to a light shade of gray (i.e., "gray90").
nrow	A single numeric that contains the number of rows to use on the graphic.
ncol	A single numeric that contains the number of columns to use on the graphic.
byrow	A single logical that indicates if the histograms should fill rows first (=TRUE or columns first (=FALSE).
iaxs	A single logical that indicates whether both axes should be plotted using $xaxs="i"$ and $yaxs="i"$ (the default) or $xaxs="r"$ and $yaxs="r"$ (what R typically does).
	Other arguments to pass through to the default hist().

Details

The formula must be of the form ~quantitative, quantitative~1, quantitative~factor, or quantitative~factor*factor2 where quantitative is the quantitative variable to construct the histograms for and factor or factor2 are factor variables that contain the levels for which separate histograms should be constructed.

If the formula is of the form ~quantitative or quantitative~1 then only a single histogram of the quantitative variable will be produced. This allows hist.formula() to be used similarly to hist() but with a data= argument.

The function produces a single (but see below) graphic that consists of a grid on which the separate histograms are printed. The rows and columns of this grid are determined to construct a plot that is as square as possible. However, the rows and columns can be set by the user with the nrow= and ncol= arguments. If the product of the number of rows and number of columns set by the user is less than the total number of histograms to be constructed then multiple pages of histograms will be produced (each requiring the user to click on the graph to go to the next graph). The x-axis of each separate histogram will be labeled identically. The default x-axis label is the name of the quantitative variable. This can be changed by the user with the xlab= argument.

94 hist formula

The default for right= is not the same as that used in hist() from **graphics**. Thus, right-open (left-closed) bins are the default.

The iaxs= argument defaulats to TRUE so that xaxs="i" and yaxs="i" are used for both axes, which eliminates the "floating" x-axis that R typically plots for histograms.

Value

A graphic is produced and nothing is returned unless formula results in only one histogram. In that case, an object of class "histogram" is returned, which is described in hist.

IFAR Chapter

3-Plotting Fundamentals.

Note

Students often need to look at the distribution of a quantitative variable separated for different levels of a categorical variable. One method for examining these distributions is with boxplot(quantitative~factor). Other methods use functions in **Lattice** and **ggplots2** but these packages have some learning 'overhead' for newbie students. The formula notation, however, is a common way in R to tell R to separate a quantitative variable by the levels of a factor. Thus, this function adds code for formulas to the generic hist function. This allows newbie students to use a common notation (i.e., formula) to easily create multiple histograms of a quantitative variable separated by the levels of a factor.

Author(s)

Derek H. Ogle, <derek@derekogle.com>, but this implementation is largely a modification of the code provided by Marc Schwartz on the R-help mailing list on 1Jun07.

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

See Also

See base hist for related functionality and multhist in plotrix for similar functionality.

```
## Using the defaults
hist(Sepal.Length~Species,data=iris)

## Add x-labels and use a pre-fix on the main labels
hist(Sepal.Length~Species,data=iris,xlab="Sepal Length (cm)",pre.main="Species==")

## Use different breaks and different y-axis limits for each graph
hist(Sepal.Length~Species,data=iris,xlab="Sepal Length (cm)",same.breaks=FALSE,same.ylim=FALSE)

## Use same but user-controlled breaks for each graph
hist(Sepal.Length~Species,data=iris,xlab="Sepal Length (cm)",breaks=seq(4,8,1))
```

hoCoef 95

```
## Use same but user-controlled maximum value for y-axis
hist(Sepal.Length~Species,data=iris,xlab="Sepal Length (cm)",ymax=30)
## Control the organization of the 'grid' of histograms
hist(Sepal.Length~Species,data=iris,xlab="Sepal Length (cm)",nrow=1,ncol=3)
## Use right=FALSE & freq=FALSE arguments to demonstrate sending an argument used by base hist()
hist(Sepal.Length~Species,data=iris,xlab="Sepal Length (cm)",right=FALSE,freq=FALSE,ymax=2)
## Add a junk variable to the iris data set to show two factors on RHS
iris$junk <- factor(sample(c("A","B"),nrow(iris),replace=TRUE))</pre>
hist(Sepal.Length~Species*junk,data=iris,xlab="Sepal Length (cm)")
## Single histogram without grouping using formula notation
hist(~Sepal.Length,data=iris,xlab="Sepal Length (cm)")
## Single histogram with "axis correction" turned off (compare to previous)
hist(~Sepal.Length,data=iris,xlab="Sepal Length (cm)",iaxs=FALSE)
## Single histogram with "axis correction", testing xlim and ylim
hist(~Sepal.Length,data=iris,xlab="Sepal Length (cm)",xlim=c(3.8,8.2),ylim=c(0,35))
hist(~Sepal.Length,data=iris,xlab="Sepal Length (cm)",xlim=c(3.8,8.2),ymax=35)
```

hoCoef

Performs a hypothesis test that a linear model parameter is equal to a specific value.

Description

Performs a hypothesis test that a linear model parameter is equal to a specific value. Useful for testing that a parameter is equal to a value other than 0.

Usage

```
hoCoef(object, term = 2, bo = 0, alt = c("two.sided", "less", "greater"))
```

Arguments

object	A 1m object.
term	A single numeric that indicates which term in the model to use in the hypothesis test.
bo	The null hypothesized parameter value.
alt	A string that identifies the "direction" of the alternative hypothesis. The strings may be "less" for a "less than" alternative, "greater" for a "greater than" alternative, or "two.sided" (DEFAULT) for a "not equals" alternative.

96 hyperCI

Details

The "direction" of the alternative hypothesis is identified by a string in the alt argument.

If the lm object is from a simple linear regression with an intercept then term=1 will use the intercept and term=2 will use the slope in the hypothesis test.

Value

A matrix that contains the term number, hypothesized value, parameter estimate, standard error of the parameter estimate, t test statistic, degrees-of-freedom, and corresponding p-value.

Author(s)

```
Derek H. Ogle, <derek@derekogle.com>
```

See Also

```
htest.nlsBoot.
```

Examples

```
data(Mirex)
# Simple linear regression test HA:slope!=0.1
lm1 <- lm(mirex~weight, data=Mirex)
hoCoef(lm1,2,0.1)</pre>
```

hyperCI

Confidence interval for population size (N) in hypergeometric distribution.

Description

Computes a confidence interval for population size (N) in hypergeometric distribution.

Usage

```
hyperCI(M, n, m, conf.level = 0.95)
```

Arguments

M	Number of successes in the population.
n	Number of observations in the sample.
m	Number of observed successes in the sample.
conf.level	Level of confidence to use for constructing confidence intervals (default is 0.95).

jolly 97

Details

This is an inefficient brute-force algorithm. The algorithm computes the conf.level range of possible values for m, as if it was unknown, for a large range of values of N. It then finds all possible values of N for which m was in the conf.level range. The smallest and largest values of N for which m was in the conf.level range are the CI endpoints.

Value

A 1x2 matrix that contains the lower and upper confidence interval bounds.

Note

This algorithm is experimental at this point.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

Examples

```
hyperCI(50,25,10)
```

jolly

Jolly-Seber analysis from multiple mark-recapture events from an open population.

Description

This function takes the two parts of a Method B table and uses the Jolly-Seber method to estimate the population size at each possible sample period and the apparent survival rate and number of additional individuals added to the population between possible sample periods. This method assumes that the population is open.

Usage

```
jolly(...)
mrOpen(mb.top, mb.bot = NULL, type = c("Jolly", "Manly"),
  conf.level = 0.95, phi.full = TRUE)

## S3 method for class 'mrOpen'
summary(object, parm = c("N", "phi", "B", "M"),
  verbose = FALSE, ...)

## S3 method for class 'mrOpen'
confint(object, parm = c("N", "phi", "B"), level = NULL,
  conf.level = NULL, verbose = FALSE, ...)
```

98 jolly

Arguments

A matrix that contains the "top" of the Method B table (i.e., a contingency table mb.top of capture sample (columns) and last seen sample (rows)) or an object of class CapHist from capHistSum. See details. mb.bot A data frame that contains the "bottom" of the Method B table (i.e., the number of marked fish in the sample (m), the number of unmarked fish in the sample (u), the total number of fish in the sample (n), and the number of marked fish returned to the population following the sample (R)). A string that indicates whether the large sample (normal theory) method of Jolly type (type="Jolly") or the "arbitrary" method of Manly (type="Manly") should be used to construct confidence intervals. conf.level A single numeric that indicates the level of confidence to use for constructing confidence intervals (default is 0.95). See details. phi.full A logical that indicates whether the standard error for phi should include only sampling variability (phi.full=FALSE) or sampling and individual variability (phi.full=TRUE,default). object An object from mr0pen (i.e., of class mr0pen). A string that identifies the model parameters for which to return summaries or parm confidence intervals. By default, all parameters are returned. verbose A logical that indicates if the observables and other notes should be printed in summary and if the type of confidence interval used should be printed in confint. See details. level Same as conf. level but used for compatability with generic confint function. Additional arguments for methods.

Details

jolly is just a convenience wrapper that produces the exact same results as mrOpen.

If mb. top contains an object from the capHistSum function then mb. bot can be left missing. In this case, the function will extract the needed data from the methodB. top and methodB. bot portions of the CapHist class object.

If mb. top is a matrix then it must be square, must have non-negative and no NA values in the upper triangle, and all NA values on the lower triangle and diagonal. If mb. bot is a matrix then it must have four rows named m, u, n, and R (see capHistSum for definitions), all values must be non-NA, and the first value of m must be 0. The last value of R can either be 0 or some positive number (it is ultimately ignored in all calculations).

All parameter estimates are performed using equations 4.6-4.9 from Pollock et al (1990) and from page 204 in Seber 2002. If type="Jolly" then all standard errors (square root of the variances) are from equations 4.11, 4.12, and 4.14 in Pollock et al. (1990) (these are different than those in Seber (2002) ... see Pollock et al.'s note on page 21). If type="Jolly" and phi.full=TRUE then the full variance for the phi parameter is given as in eqn 4.18 in Pollock et al. (1990), otherwise eqn 4.13 from Pollock et al. (1990) is used. When type="Jolly" the confidence interval are produced using normal theory (i.e., estimate +/- z*SE). If type="Manly" then the confidence intervals for N and phi (none will be produced for B) are constructed using the methods of Manly (1984) and as described in 2.24-2.33 of Krebs (1989). No standard errors are returned when type="Manly".

The summary function returns estimates of M, N, phi, B, and their associated standard errors and, if verbose=TRUE the intermediate calculations of "observables" from the data – n, m, R, r, and z.

The level of confidence is not set in the confint function, in contrast to most confint functions. Rather the confidence level is set in the main mrOpen function.

Value

A list with the following items:

- df A data frame that contains observable summaries from the data and estimates of the number of extant marked fish (M), population size for each possible sample period (N), apparent survival rate between each possible pair of sample periods (phi), and the number of additional individuals added to the population between each possible pair of sample periods (B). In addition to the estimates, values of the standard errors and the lower and upper confidence interval bounds for each parameter are provided (however, see the details above).
- type The provided type of confidence intervals that was used.
- phi.full The provided logical that indicates the type of standard error for phi that was used.
- conf.level The provided level of confidence that was used.

Testing

The formulas have been triple-checked against formulas in Pollock et al. (1990), Manly (1984), and Seber (2002).

The results for the CutthroatAL data file (as analyzed in the example) was compared to results from the JOLLY program available at http://www.mbr-pwrc.usgs.gov/software/jolly.html. The r and z values matched, all M and N estimates match at one decimal place, all phi are within 0.001, and all B are within 0.7. The SE match for M except for two estimates that are within 0.1, match for N except for one estimate that is within 0.1, are within 0.001 for phi, and are within 1.3 for B (except for for the first estimate which is dramatically off).

The results of mr0pen related to Table 4.4 of Pollock et al. (1990) match (to one decimal place) except for three estimates that are within 0.1% for N, match (to two decimal places) for phi except for where Pollock set phi>1 to phi=1, match for B except for Pollock set B<0 to B=0. The SE match (to two decimal places) for N except for N15 (which is within 0.5, <5%), match (to three decimal places) for phi except for phi15 (which is within 0.001, <0.5%), match (to two decimal places) for B except for B17 and B20 which are within 0.2 (<0.2%)

All point estimates of M, N, phi, and B and the SE of phi match the results in Table 2.3 of Krebs (1989) (within minimal rounding error for a very small number of results). The SE of N results are not close to those of Krebs (1989) (who does not provide a formula for SE so the discrepancy cannot be explored). The SE of B results match those of Krebs (1989) for 5 of the 8 values and are within 5% for 2 of the other 3 values (the last estimate is off by 27%).

For comparing to Jolly's data as presented in Tables 5.1 and 5.2 of Seber (2002), M was within 4 (less than 1.5%), N was within 3% (except N2 which was within 9%), phi was within 0.01 (less than 1.5

IFAR Chapter

9-Abundance from Capture-Recapture Data and 11-Mortality.

jolly

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Jolly, G.M. 1965. Explicit estimates from capture-recapture data with both death and immigration – stochastic model. Biometrika, 52:225-247.

Krebs, C.J. 1989. Ecological Methodology. Harper & Row Publishers, New York.

Leslie, P.H. and D. Chitty. 1951. The estimation of population parameters from data obtained by means of the capture-recapture method. I. The maximum likelihood equations for estimating the death-rate. Biometrika, 38:269-292.

Manly, B.F.J. 1984. Obtaining confidence limits on parameters of the Jolly-Seber model for capture-recapture data. Biometrics, 40:749-758.

Pollock, K.H., J.D. Nichols, C. Brownie, and J.E. Hines. 1991. Statistical inference for capture-recapture experiments. Wildlife Monographs, 107:1-97.

Seber, G.A.F. 1965. A note on the multiple recapture census. Biometrika 52:249-259.

Seber, G.A.F. 2002. The Estimation of Animal Abundance. Edward Arnold, second edition (reprinted).

See Also

```
capHistSum, mrClosed
```

```
## First example -- capture histories summarized with capHistSum()
data(CutthroatAL)
ch1 <- capHistSum(CutthroatAL,cols2use=-1) # ignore first column of fish ID
ex1 <- mrOpen(ch1)
summary(ex1)
summary(ex1,verbose=TRUE)
summary(ex1,parm="N")
summary(ex1,parm=c("N","phi"))
confint(ex1)
confint(ex1,parm="N")
confint(ex1,parm=c("N","phi"))
confint(ex1,verbose=TRUE)
## Second example - Jolly's data -- summarized data entered "by hand"
s1 < - rep(NA, 13)
s2 <- c(10, rep(NA, 12))
s3 < -c(3,34,rep(NA,11))
s4 <- c(5,18,33,rep(NA,10))
s5 < c(2,8,13,30,rep(NA,9))
s6 < c(2,4,8,20,43,rep(NA,8))
s7 <- c(1,6,5,10,34,56,rep(NA,7))
s8 \leftarrow c(0,4,0,3,14,19,46,rep(NA,6))
s9 \leftarrow c(0,2,4,2,11,12,28,51,rep(NA,5))
```

kCounts 101

```
s10 < c(0,0,1,2,3,5,17,22,34,rep(NA,4))
s11 < c(1,2,3,1,0,4,8,12,16,30,rep(NA,3))
s12 <- c(0,1,3,1,1,2,7,4,11,16,26,NA,NA)
s13 \leftarrow c(0,1,0,2,3,3,2,10,9,12,18,35,NA)
jolly.top <- cbind(s1, s2, s3, s4, s5, s6, s7, s8, s9, s10, s11, s12, s13)
n \leftarrow c(54,146,169,209,220,209,250,176,172,127,123,120,142)
R \leftarrow c(54,143,164,202,214,207,243,175,169,126,120,120,0)
m \leftarrow c(0,10,37,56,53,77,112,86,110,84,77,72,95)
u <- n-m
jolly.bot <- rbind(m,u,n,R)</pre>
ex2 <- mrOpen(jolly.top,jolly.bot)</pre>
summary(ex2,verbose=TRUE)
confint(ex2,verbose=TRUE)
ex3 <- mrOpen(jolly.top,jolly.bot,type="Manly")</pre>
summary(ex3,verbose=TRUE)
confint(ex3,verbose=TRUE)
## demonstrate use of jolly()
ex3a <- jolly(jolly.top,jolly.bot)</pre>
```

kCounts

Specific utilities for use in a knitr document.

Description

Specific utilities for pretty printing various items in a knitr document.

Usage

```
kCounts(value, capitalize = FALSE)
kPvalue(value, digits = 4, include.p = TRUE, latex = TRUE)
purl2(file, out.dir = NULL, newname = NULL, topnotes = NULL,
    moreItems = NULL, blanks = c("extra", "all", "none"), timestamp = TRUE,
    ...)
reproInfo(out = c("r", "markdown", "latex"), rqrdPkgs = NULL,
    elapsed = NULL, width = 0.95 * getOption("width"), addTOC = TRUE,
    newpage = FALSE, links = NULL, closeGraphics = TRUE)
```

Arguments

value A single numeric count or p-value.

capitalize A logical that indicates if the returned words should be capitalized or not (the

default).

102 kCounts

digits	Number of decimal places to round the values to.
include.p	A logical that indicates whether the result should be a character string with "p=" appended to the numerical result.
latex	A logical that indicates whether the resultant p-value string should be contained within dollar signs to form a latex formula.
file	A string that contains the root name of the .RNW file. This will also be the name of the resultant purled file with .R appended.
out.dir	A string that indicates the directory structure in which the purled file should be located. This should not have a forward slash at the end.
newname	A string for the output filename (without the extension) from purl2.
topnotes	A character vector of lines to be added to the top of the output file. Each value in the vector will be placed on a single line at the top of the output file.
moreItems	A string that contains additional words that when found in the purled file will result in the entire line with those words to be deleted.
blanks	A string that indicates if blank lines should be removed. If blanks="all" then all blank lines will be removed. If blanks="extra" then only "extra" blanks lines will be removed (i.e., one blank line will be left where there was originally more than one blank line).
timestamp	A logical that indicates whether a timestamp comment should be appended to the bottom of the script created by purl2.
out	A string that indicates the type of output from reproInfo – Markdown, LaTeX, or simple R code.
rqrdPkgs	A string vector that contains packages that are required for the vignette and for which all dependencies should be found.
elapsed	A numeric, usually from proc.time, that is the time required to run the vignette. If NULL then this output will not be used. See the note below.
width	A numeric that indicates the width to use for wrapping the reproducibility information when out="r".
addT0C	A logical that indicates whether or not a table of contents entry for the reproducibity section should be added to the LaTeX output. Used only if Rout="latex"
newpage	A logical that indicates whether or not the reproduciility information should begin on a new page. Used only if Rout="latex"
links	A named character vector that will add a links bullet to the reproducibility information. The names will be shown and the values are the links. Used only if Rout="markdown.
closeGraphics	A logical that indicates whether the graphics device should be closed or not.
	Additional arguments for the original pur1.

Details

- kCounts is used to convert numeric numbers to 'word' numbers in a sentence.
- kPvalue is used to print 'pretty' p-values.
- purl2 is used to create a modified (see below) Stangled or purled script.
- reproInfo is used to print 'reproducibility information' for the document.

kCounts 103

Value

• kCounts returns a numeric value if the count is less than zero or greater than ten and returns a character string of the number 'name'. See the examples.

- kPvalue returns a character string of the supplied p-value rounded to the requested number of digits or a character string that indicates what the p-value is less than the value with a '5' in the digits+1 place. See the examples.
- purl2 is a modification of purl from **knitr** that creates a file with the same name as file but with lines removed that contain certain words (those found in ItemsToRemove and moreItems).
- reproInfo returns Markdown, LaTeX, or R code that prints "reproducibility information" at the bottom of the knitted document.

Note

In reproInfo, elapsed can be used to print the time it took to process the document by sending the elasped time for processing to this argument. The simplest way to get an approximate elapsed time is to put st <- proc.time() very early (first line?) in your knitr code, put et <- proc.time()-st very late in your knitr code (i.e., just prior to reproInfo), and then used elapsed=et["user.self"]+et["sys.self"] in reproInfo.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

See Also

See formatC for functionality similar to kPvalue. See purl and knit in **knitr** for functionality similar to purl2.

```
kCounts(7)
kCounts(17)
kCounts(0)
kCounts(-6)
kCounts(3,capitalize=TRUE)

kPvalue(0.123456789)
kPvalue(0.000123456)
kPvalue(0.000012345)
kPvalue(0.000012345,include.p=FALSE)
kPvalue(0.000012345,include.p=FALSE,latex=FALSE)
```

104 ksTest

l _c 1	Tac+
NO I	CSL

Kolmogorov-Smirnov Tests.

Description

Performs a one- or two-sample Kolmogorov-Smirnov test. Includes the option to perform the two-sample test using the formula notation.

Usage

```
ksTest(x, ...)
## Default S3 method:
ksTest(x, y, ..., alternative = c("two.sided", "less",
    "greater"), exact = NULL)
## S3 method for class 'formula'
ksTest(x, data = NULL, ..., alternative = c("two.sided",
    "less", "greater"), exact = NULL)
```

Arguments

X	A numeric vector of data values or a formula (see details).
У	A numeric vector of data values, a character string naming a cumulative distribution function, or an actual cumulative distribution function. See ks.test.
alternative	A string that indicates the alternative hypothesis. See ks.test.
exact	NULL or a logical that indicates whether an exact p-value should be computed. See ks.test. Not available if ties are present, nor for the one-sided two-sample case.
data	A data frame that contains the variables in the formula for x.
	Parameters of the distribution specified (as a character string) by y.

Details

This is exactly ks.test except that a formula may be used for the two-sample situation. The default version is simply a pass through to ks.test. See ks.test for more details.

Value

```
See ks. test.
```

Author(s)

Derek H. Ogle, <derek@derekogle.com>

lagratio 105

See Also

```
ks.test.
```

Examples

```
## see ks.test for other examples
x <- rnorm(50)
y <- runif(30)
df <- data.frame(dat=c(x,y),grp=rep(c("X","Y"),c(50,30)))
## one-sample (from ks.test) still works
ksTest(x+2, "pgamma", 3, 2)
ks.test(x+2, "pgamma", 3, 2)
## first two-sample example in ?ks.test
ksTest(x,y)
ks.test(x,y)
## same as above but using data.frame and formula
ksTest(dat~grp,data=df)</pre>
```

lagratio

Ratio of lagged observations.

Description

Computes the ratio of lagged observations in a vector.

Usage

```
lagratio(x, lag = 1L, recursion = 1L, differences = recursion,
  direction = c("backward", "forward"), ...)
```

Arguments

X	A numeric vector or matrix.
lag	An integer representing the lag 'distance'.
recursion	An integeer that indicates the level of recursion for the calculations. A 1 will simply compute the ratios. A 2, for example, will compute the ratios, save the result, and then compute the ratios of the results using the same lag. See examples.
differences	Same as recursion. Used for symmetry with diff.
direction	A string that indicates the direction of calculation. A "backward" induicates that 'latter' values are divided by 'former' values. A "forward" induicates that 'former' values are divided by 'latter' values. See examples.

... Additional arguments to diff().

106 lencat

Details

This function behaves similarly to diff() except that it returns a vector or matrix of ratios rather than differences.

Value

A vector or matrix of lagged ratios.

Author(s)

```
Derek H. Ogle, <derek@derekogle.com>
```

See Also

diff

Examples

```
## Backward lagged ratios
# no recursion
lagratio(1:10,1)
lagratio(1:10,2)
# with recursion
lagratio(1:10,1,2)
lagratio(1:10,2,2)

## Forward lagged ratios
# no recursion
lagratio(10:1,1,direction="forward")
lagratio(10:1,2,direction="forward")
# with recursion
lagratio(10:1,1,2,direction="forward")
lagratio(10:1,2,2,direction="forward")
lagratio(10:1,2,2,direction="forward")
```

lencat

Constructs length class/category variable.

Description

Constructs a vector that contains the length class or category to which an individual belongs. Optionally, that vector can be appened to the original data frame.

Usage

```
lencat(x, ...)
## Default S3 method:
lencat(x, w = 1, breaks = NULL, startcat = NULL,
```

lencat 107

```
right = FALSE, use.names = FALSE, as.fact = use.names,
  droplevels = drop.levels, drop.levels = FALSE, ...)

## S3 method for class 'formula'
lencat(x, data, w = 1, breaks = NULL, startcat = NULL,
  right = FALSE, use.names = FALSE, as.fact = use.names,
  droplevels = drop.levels, drop.levels = FALSE, vname = NULL, ...)
```

Arguments

X	A numeric vector that contains the length measurements or a formula of the form ~x where "x" generically represents a variable in data that contains length measurements. This formula can only contain one variable.
W	A single numeric that indicates the width of length categories to create.
breaks	A numeric vector of lower values for the break points of the length categories.
startcat	A single numeric that indicates the beginning of the first length category.
right	A logical that indicates if the intervals should be closed on the right (and open on the left) or vice versa.
use.names	A logical that indicates whether the names for the values in breaks should be used for the levels in the new variable. Will throw a warning and then use default levels if TRUE but names(breaks) is NULL.
as.fact	A logical that indicates that the new variable should be returned as a factor (=TRUE) or not (=FALSE; default).
droplevels,drop.levels	
	A logical that indicates that the new variable should retain all levels indicated in breaks (=FALSE; default) or not. Ignored if as.fact=FALSE.
data	A data frame that minimally contains the length measurements given in the variable in the formula.
vname	A string that contains the name for the new length class variable.
	Not implemented.

Details

If breaks is non-NULL then w and startcat will be ignored. The vector of values in breaks should begin with a value smaller than the minimum observed value and end with a value larger than the maximum observed value. If the lowest break value is larger than the minimum observed value then an error will occur. If the largest break value is smaller than the maximum observed value then an additional break value larger than the maximum observed value will be added to breaks (and a warning will be sent). The values in breaks do not have to be equally spaced.

If breaks=NULL (the default) then the value in w is used to create equally spaced categories. The start of the length categories can be set with startcat. However, if startcat=NULL (the default) the the length categories will begin with the first value less than the minimum observed value when "rounded" by w. For example, if the minimum observed value is 67 then the first length category will be 65 if w=5, 60 if w=10, 50 if w=25, and 50 if w=50. The length categories will continue from this starting value by values of w until a category value is greater than the largest observed value in v. The length categories are left-inclusive and right-exclusive by default (i.e., right=FALSE).

108 lencat

The number in the startcat argument should be less than the smallest value in x. Additionally, the number of decimals in startcat should not be more than the number of decimals in w (e.g., startcat=0.4 and w=1 will result in an error).

One may want to convert apparent numeric values to factor values if some of the length categories are missing (e.g., if factor values are used, for example, then tables of the length categories values will have values for all length categories; i.e., it will have zeroes for the length categories that are missing). The numeric values can be converted to factors by including as . fact. See the "real data" example.

The observed values in the x should be rounded to the appropriate number of decimals to avoid misplacement of individuals into incorrect length categories due to issues with machine-precision (see discussion in all.equal.)

Value

If the formula version of the function is used then a data frame will be returned with the a new variable, named as in vname or LCat if no name is given by the user, appended to the original data frame. If the default version of the function is used then a single vector will be returned. The returned values will be numeric unless breaks is named and use.names=TRUE or if as.fact=TRUE.

IFAR Chapter

2-Data Manipulation.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

```
# Create random lengths measured to nearest 0.1 unit
df1 <- data.frame(len=round(runif(50,0.1,9.9),1))

# Create length categories by 0.1 unit
df1$LCat1 <- lencat(df1$len,w=0.1)
xtabs(~LCat1,data=df1)

# length categories by 0.2 units
df1$LCat2 <- lencat(df1$len,w=0.2)
xtabs(~LCat2,data=df1)

# length categories by 0.2 units starting at 0.1
df1$LCat3 <- lencat(df1$len,w=0.2,startcat=0.1)
xtabs(~LCat3,data=df1)

# length categories as set by breaks
df1$LCat4 <- lencat(df1$len,breaks=c(0,2,4,7,10))</pre>
```

lencat 109

```
xtabs(~LCat4,data=df1)
## A Second example
# random lengths measured to nearest unit
df2 <- data.frame(len=round(runif(50,10,117),0))</pre>
# length categories by 5 units
df2$LCat1 <- lencat(df2$len,w=5)</pre>
xtabs(~LCat1,data=df2)
# length categories by 5 units starting at 7
df2$LCat2 <- lencat(df2$len,w=5,startcat=7)</pre>
xtabs(~LCat2,data=df2)
# length categories by 10 units
df2$LCat3 <- lencat(df2$len,w=10)</pre>
xtabs(~LCat3,data=df2)
# length categories by 10 units starting at 5
df2$LCat4 <- lencat(df2$len,w=10,startcat=5)</pre>
xtabs(~LCat4,data=df2)
# length categories as set by breaks
df2$LCat5 <- lencat(df2$len,breaks=c(5,50,75,150))</pre>
xtabs(~LCat5,data=df2)
## A Third example
# random lengths measured to nearest 0.1 unit
df3 <- data.frame(len=round(runif(50,10,117),1))</pre>
# length categories by 5 units
df3$LCat1 <- lencat(df3$len,w=5)</pre>
xtabs(~LCat1,data=df3)
## A Fourth example
# random lengths measured to nearest 0.01 unit
df4 <- data.frame(len=round(runif(50,0.1,9.9),2))</pre>
# length categories by 0.1 unit
df4$LCat1 <- lencat(df4$len,w=0.1)
xtabs(~LCat1,data=df4)
\# length categories by 0.1 unit, but without missing categories
df4$LCat2 <- lencat(df4$len,w=0.1,as.fact=TRUE)</pre>
xtabs(~LCat2,data=df4)
# length categories by 2 unit
df4$LCat3 <- lencat(df4$len,w=2)
xtabs(~LCat3,data=df4)
## A Fifth example -- with real data
data(SMBassWB)
# remove variables with "anu" and "radcap" just for simplicity
```

110 lencat

```
smb1 \leftarrow smb2 \leftarrow SMBassWB[,-c(8:20)]
# 10 mm length classes - in default LCat variable
smb1$LCat10 <- lencat(smb1$lencap,w=10)</pre>
head(smb1)
xtabs(~LCat10,data=smb1)
# Same as previous but returned as a factor so that levels with no fish are still seen
smb1$LCat10A <- lencat(smb1$lencap,w=10,as.fact=TRUE)</pre>
head(smb1)
xtabs(~LCat10A, data=smb1)
# Same as previous but returned as a factor with unused levels dropped
smb1$LCat10B <- lencat(smb1$lencap,w=10,as.fact=TRUE,droplevels=TRUE)</pre>
head(smb1)
xtabs(~LCat10B, data=smb1)
# 25 mm length classes - in custom variable name
smb1$LCat25 <- lencat(smb1$lencap,w=25)</pre>
head(smb1)
xtabs(~LCat25,data=smb1)
# using values from psdVal for Smallmouth Bass
smb1$PSDCat1 <- lencat(smb1$lencap,breaks=psdVal("Smallmouth Bass"))</pre>
head(smb1)
xtabs(~PSDCat1,data=smb1)
# add category names
smb1$PSDCat2 <- lencat(smb1$lencap,breaks=psdVal("Smallmouth Bass"),use.names=TRUE)</pre>
head(smb1)
xtabs(~PSDCat2,data=smb1)
# same as above but drop the unused levels
smb1$PSDCat2A <- lencat(smb1$lencap,breaks=psdVal("Smallmouth Bass"),</pre>
                         use.names=TRUE,droplevels=TRUE)
head(smb1)
xtabs(~PSDCat2A,data=smb1)
str(smb1)
# same as above but not returned as a factor (returned as a character)
smb1$PSDcat2B <- lencat(smb1$lencap,breaks=psdVal("Smallmouth Bass"),</pre>
                         use.names=TRUE,as.fact=FALSE)
str(smb1)
## A Sixth example -- similar to the fifth example but using the formula notation
# 10 mm length classes - in default LCat variable
smb2 <- lencat(~lencap,data=smb2,w=10)</pre>
head(smb2)
# 25 mm length classes - in custom variable name
smb2 <- lencat(~lencap,data=smb2,w=25,vname="LenCat25")</pre>
head(smb2)
```

lenFreqExpand 111

lenFreqExpand

Expands a length frequency based on a subsample.

Description

Creates a vector of lengths for the individuals not measured based on the lengths measured in a subsample of individuals.

Usage

```
lenFreqExpand(...)
expandLenFreq(x, w, additional, startcat = NULL, total = additional +
  length(x), decimals = decs$wdec, show.summary = TRUE, ...)
```

Arguments

X	A numeric vector of length measurements.	
W	A number that indicates the width of length classes to create.	
additional	The number of individuals that were not measured in the sample (for which measurements should be determined).	
startcat	A number that indicates the beginning of the first length-class.	
total	The total number of individuals in the sample (including those that were measured in the subsample).	
decimals	A number that indicates the number of decimals used in the output vector of estimated lengths.	
show.summary	A logical that indicates whether a summary of the process should be shown at the end.	
	Optional arguments to be passed to lencat.	

112 lenFreqExpand

Details

Creates a vector of lengths for the individuals not measured based on the lengths measured in a subsample of individuals. Length categories are created first that begin with the value in startcat (or the minimum observed value by default) and continue by values of w until a category value greater than the largest observed length in x. Categories of different widths are not allowed.

The resulting "expanded" lengths are created by allocating individuals to each length class based on the proportion of measured individuals in the subsample in that length class. Individuals within a length class are then assigned a specific length within that length class based on a uniform distribution. Because the expanded number of individuals in a length class is rounded down based on the measured number per length class, not all individuals will initially be assigned a length value. The remaining individuals are assigned to a length class randomly according to weights based on the proportion of individuals in the measured length classes. Finally, these individuals are randomly assigned a specific length within the respective length class from a uniform distribution, same as above.

The resulting length assignments are rounded to the number of decimals shown in decimal. If decimals is not set by the user then it will default to the same number of decimals shown in the w value. Care is taken to make sure that the rounded result will not pass out of the given length category (i.e., will not be allowed to round up to the next length category). Generally speaking, one will want to use more decimals then is shown in w. For example, one may want to create length categories with a width of 1 inch (i.e., w=1) but have the results recorded as if measured to within 0.1 inch (i.e., decimals=1).

Value

Returns a vector that consists of measurements for the non-measured individuals in the entire sample.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

See Also

See expandCounts for expanding more than just lengths or expanding lengths when there is a known number in each length bin. See lencat for creating length bins.

```
## Set the random seed for reproducibility
set.seed(15343437)

## First example
# random lengths measured to nearest 0.1 unit -- values in a vector
len1 <- round(runif(50,0.1,9.9),1)
# assignment of integer lengths to 110 non-measured individuals
new.len1a <- expandLenFreq(len1,w=1,total=160)
new.len1a
# assignment of lengths to 0.1 to 110 non-measured individuals
new.len1b <- expandLenFreq(len1,w=1,total=160,decimals=1)</pre>
```

logbtcf 113

```
new.len1b
## Second example -- if values are in a data.frame
# random lengths measured to nearest 0.1 unit
len2 <- data.frame(len=round(runif(50,10,117),1))</pre>
# assignment of lengths to 0.1 for 140 non-measured indivs
new.len2a <- expandLenFreq(len2$len,w=10,total=190,decimals=1)</pre>
new.len2a
## Third example
# hypothetically measured lengths
len \leftarrow c(6.7,6.9,7.3,7.4,7.5,8.2,8.7,8.9)
# find lengths for unmeasured fish assuming a total of 30
newlen1 <- expandLenFreq(len,w=0.5,total=30,decimals=1)</pre>
newlen1
# set a starting length category
newlen2 <- expandLenFreq(len,w=0.5,startcat=6.2,total=30,decimals=1)</pre>
newlen2
```

logbtcf

Constructs the correction-factor used when back-transforming log-transformed values.

Description

Constructs the correction-factor used when back-transforming log-transformed values according to the method in Sprugel (1983). Sprugel's main formula – exp((syx^2)/2) – is used when syx is estimated for natural log transformed data. He noted that a formula for any based could be constructed by multiplying the syx term by log_e(base) to give exp(((log_e(base)*syx)^2)/2). This more general formula is implemented in this function (where, of course, if the base is exp(1) then the general formula reduces to the original specific formula.)

Usage

```
logbtcf(obj, base = exp(1))
```

Arguments

obj An object from 1m.

base A single numerica value that indicates the base of the logarithm used.

Value

A numeric value that is the correction factor according to Sprugel (1983).

Author(s)

Derek H. Ogle, <derek@derekogle.com>

114 lwCompPreds

References

Sprugel, D.G. 1983. Correcting for bias in log-transformed allometric equations. Ecology 64:209-210

Examples

```
# toy data
df <- data.frame(y=rlnorm(10),x=rlnorm(10))</pre>
df$logey <- log(df$y)</pre>
df$log10y <- log10(df$y)
df$logex <- log(df$x)
df$log10x <- log10(df$x)
# model and predictions on loge scale
lme <- lm(logey~logex,data=df)</pre>
( ploge <- predict(lme,data.frame(logex=log(10))) )</pre>
( pe <- exp(ploge) )
( cfe <- logbtcf(lme) )</pre>
( cpe <- cfe*pe )
# model and predictions on log10 scale
lm10 \leftarrow lm(log10y\sim log10x, data=df)
plog10 <- predict(lm10,data.frame(log10x=log10(10)))</pre>
p10 <- 10^(plog10)
( cf10 <- logbtcf(lm10,10) )</pre>
(cp10 < -cf10*p10)
# cfe and cf10, cpe and cp10 should be equal
all.equal(cfe,cf10)
all.equal(cpe,cp10)
```

1wCompPreds

Constructs plots of predicted weights at given lengths among different groups.

Description

Constructs plots of predicted weights at given lengths among different groups. These plots allow the user to explore differences in predicted weights at a variety of lengths when the weight-length relationship is not the same across a variety of groups.

Usage

```
lwCompPreds(object, lens = NULL, qlens = c(0.05, 0.25, 0.5, 0.75, 0.95),
  qlens.dec = 1, base = exp(1), interval = c("confidence", "prediction",
  "both"), center.value = 0, lwd = 1, connect.preds = TRUE,
  show.preds = FALSE, col.connect = "gray50", ylim = NULL,
  main.pre = "Length==", cex.main = 0.8, xlab = "Groups",
  ylab = "Predicted Weight", yaxs = "r", rows = round(sqrt(num)),
  cols = ceiling(sqrt(num)))
```

IwCompPreds 115

Arguments

object	An lm object (i.e., returned from fitting a model with lm). This model should have log(weight) as the response and log(length) as the explanatory covariate and an explanatory factor variable that describes the different groups.
lens	A numeric vector that indicates the lengths at which the weights should be predicted.
qlens	A numeric vector that indicates the quantiles of lengths at which weights should be predicted. This is ignored if lens is non-null.
qlens.dec	A single numeric that identifies the decimal place that the lengths derived from qlens should be rounded to (Default is 1).
base	A single positive numeric value that indicates the base of the logarithm used in the 1m object in object. The default is exp(1), or the value e.
interval	A single string that indicates whether to plot confidence (="confidence"), prediction (="prediction"), or both (="both") intervals.
center.value	A single numeric value that indicates the log length used if the log length data was centered when constructing object.
lwd	A single numeric that indicates the line width to be used for the confidence and prediction interval lines (if not interval="both") and the prediction connections line. If interval="both" then the width of the prediction interval will be one less than this value so that the CI and PI appear different.
connect.preds	A logical that indicates whether the predicted values should be connected with a line across groups or not.
show.preds	A logical that indicates whether the predicted values should be plotted with a point for each group or not.
col.connect	A color to use for the line that connects the predicted values (if $connect.preds=TRUE$).
ylim	A numeric vector of length two that indicates the limits of the y-axis to be used for each plot. If null then limits will be chosen for each graph individually.
main.pre	A character string to be used as a prefix for the main title. See details.
cex.main	A numeric value for the character expansion of the main title. See details.
xlab	A single string for labelling the x-axis.
ylab	A single string for labelling the y-axis.
yaxs	A single string that indicates how the y-axis is formed. See par for more details.
rows	A single numeric that contains the number of rows to use on the graphic.
cols	A single numeric that contains the number of columns to use on the graphic.
• • •	Other arguments to pass through to the plot function.

Value

None. However, a plot is produced.

IFAR Chapter

7-Weight-Length.

116 lwCompPreds

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

```
# load weight-length data
data(ChinookArg)
# add log length and weight data
ChinookArg$logtl <- log(ChinookArg$tl)</pre>
ChinookArg$logwt <- log(ChinookArg$w)</pre>
# fit model to assess equality of slopes
lm1 <- lm(logwt~logtl*loc,data=ChinookArg)</pre>
anova(lm1)
# set par so that the plots will look decent
par(mar=c(3.5,3.5,1,1), mgp=c(1.8,0.4,0), tcl=-0.2)
# show predicted weights (w/ CI) at the default quantile lengths for each year
lwCompPreds(lm1,xlab="Location")
# show predicted weights (w/ CI) at the quartile lengths for each year
lwCompPreds(lm1,xlab="Location",qlens=c(0.25,0.5,0.75))
# show predicted weights (w/ CI) at certain lengths for each year
lwCompPreds(lm1,xlab="Location",lens=c(60,90,120,150))
# show predicted weights (w/ just PI) at certain lengths for each year
lwCompPreds(lm1,xlab="Location",lens=c(60,90,120,150),interval="prediction")
# show predicted weights (w/ CI and PI) at certain lengths for each year
lwCompPreds(lm1,xlab="Location",lens=c(60,90,120,150),interval="both")
# show predicted weights (w/ CI and points at the prediction) at certain lengths for each year
lwCompPreds(lm1,xlab="Location",lens=c(60,90,120,150),show.preds=TRUE)
# show predicted weights (w/ CI but don't connect means) at certain lengths for each year
lwCompPreds(lm1,xlab="Location",lens=c(60,90,120,150),connect.preds=FALSE,show.preds=TRUE)
# fit model with centered data
mn.logtl <- mean(ChinookArg$logtl,na.rm=TRUE)</pre>
ChinookArg$clogtl <- ChinookArg$logtl-mn.logtl
lm2 <- lm(logwt~clogtl*loc,data=ChinookArg)</pre>
lwCompPreds(lm2,xlab="Location",center.value=mn.logtl)
lwCompPreds(lm2,xlab="Location",lens=c(60,90,120,150),center.value=mn.logtl)
# fit model with a different base (plot should be the same as the first example)
ChinookArg$logtl <- log10(ChinookArg$tl)</pre>
ChinookArg$logwt <- log10(ChinookArg$w)</pre>
lm1 <- lm(logwt~logtl*loc,data=ChinookArg)</pre>
lwCompPreds(lm1,base=10,xlab="Location")
if (interactive()) {
 # should give error, does not work for only a simple linear regression
 lm2 <- lm(logwt~logtl,data=ChinookArg)</pre>
 lwCompPreds(lm2)
```

mapvalues 117

```
# or a one-way ANOVA
lm3 <- lm(logwt~loc,data=ChinookArg)
lwCompPreds(lm3)
}</pre>
```

mapvalues

Replace specified values in a vector or factor with new values.

Description

Same function as in **plyr**. Imported by **FSA** and exported to minimize conflicts with other commonly used packages. See mapvalues for documentation.

Mirex

Mirex concentration, weight, capture year, and species of Lake Ontario salmon.

Description

Mirex concentration, weight, capture year, and species of Lake Ontario Coho and Chinook salmon.

Format

A data frame with 122 observations on the following 4 variables.

```
year a numeric vector of capture years.
weight a numeric vector of salmon weights (kg).
mirex a numeric vector of mirex concentration in the salmon tissue (mg/kg).
species a factor with levels chinook and coho.
```

Details

The year variable should be converted to a factor as shown in the example.

Source

actual data from Makarewicz, J.C., E.Damaske, T.W. Lewis, and M. Merner. 2003. Trend analysis reveals a recent reduction in mirex concentrations in coho (*Oncorhynchus kistuch*) and chinook (*O. tshawytscha*) salmon from Lake Ontario. Environmental Science and Technology, 37:1521-1527.

```
data(Mirex)
Mirex$year <- factor(Mirex$year)
lm1 <- lm(mirex~weight*year*species,data=Mirex)
anova(lm1)</pre>
```

Mmethods

Estimate natural mortality from a variety of empirical methods.

Description

Several methods can be used to estimated natural mortality (M) from other types of data, including parameters from the von Bertalanffy growth equation, maximum age, and temperature. These relationships have been developed from meta-analyses of a large number of populations. Several of these methods are implemented in this function.

Usage

```
Mmethods(what = c("all", "tmax", "K", "Hoenig", "Pauly"))

metaM(method = Mmethods(), justM = TRUE, tmax = NULL, K = NULL,
    Linf = NULL, t0 = NULL, b = NULL, L = NULL, T = NULL, t50 = NULL,
    Winf = NULL)

## S3 method for class 'metaM'
print(x, digits = 4, ...)
```

Arguments

what	A string that indicates what grouping of methods to return. Defaults to returning all methods.
method	A string that indicates which method or equation to use. See details.
justM	A logical that indicates whether just the estimate of M (TRUE; Default) or a more descriptive list should be returned.
tmax	The maximum age for the population of fish.
K	The Brody growth coefficident from the fit of the von Bertalanffy growth function.
Linf	The asymptotic mean length (cm) from the fit of the von Bertalanffy growth function.
t0	The x-intercept from the fit of the von Bertalanffy growth function.
b	The exponent from the weight-length relationship (slope from the logW-logL relationship).
L	The body length of the fish (cm).
Т	The temperature experienced by the fish (C).
t50	The age (time) when half the fish in the population are mature.
Winf	The asymptotic mean weight (g) from the fit of the von Bertalanffy growth function.
X	A metaM object returned from metaM when justM=FALSE.
digits	A numeric that controls the number of digits printed for the estimate of M.
• • •	Additional arguments for methods. Not implemented.

Details

One of several methods is chosen with method. The available methods can be seen with Mmethods() and are listed below with a brief description of where the equation came from. The sources (listed below) should be consulted for more specific information.

- method="tmax1": The "one-parameter tmax equation" from the first line of Table 3 in Then et al. (2015). This method was the preferred method suggested by Then et al. (2015). Requires only tmax.
- method="PaulyLNoT": The "modified Pauly length equation" as described on the sixth line of Table 3 in Then et al. (2015). Then et al. (2015) suggested that this is the preferred model if maximum age (tmax) information was not available. Requires K and Linf.
- method="PaulyL": The "Pauly (1980) equation using fish lengths" from his equation 11. This is the most commonly used method in the literature. Note that Pauly used common logarithms as used here but the model is often presented in other sources with natural logarithms. Requires K, Linf, and T.
- method="PaulyW": The "Pauly (1980) equation for weights" from his equation 10. Requires K, Winf, and T.
- method="Hoeing0", method="Hoeing0F", method="Hoeing0M", method="Hoeing0C": The original "Hoenig (1983) composite", "fish", "mollusc", and "cetacean" (fit with OLS) equations from the second column on page 899 of Hoenig (1983). Requires only tmax.
- method="Hoeing02", method="Hoeing02F", method="Hoeing02M", method="Hoeing02C": The original "Hoenig (1983) composite", "fish", "mollusc", and "cetacean" (fit with Geometric Mean Regression) equations from the second column on page 537 of Kenchington (2014). Requires only tmax.
- method="HoenigLM": The "modified Hoenig equation derived with a linear model" as described in Then et al. (2015) on the second line of Table 3. Requires only tmax.
- method="HoenigNLS": The "modified Hoenig equation derived with a non-linear model" as described in Then et al. (2015) on the third line of Table 3. Requires only tmax.
- method="HewittHoenig": The "Hewitt and Hoenig (2005) equation" from their equation 8. Requires only tmax.
- method="K1": The "one-parameter K equation" from the fourth line of Table 3 in Then et al. (2015). Requires only K.
- method="K2": The "two-parameter K equation" from the fifth line of Table 3 in Then et al. (2015). Requires only K.
- method="JensenK1": The "Jensen (1996) one-parameter K equation". Requires only K.
- method="JensenK2": The "Jensen (2001) two-parameter K equation" from their equation 8. Requires only K.
- method="Gislason": The "Gislason et al. (2010) equation" from their equation 2. Requires K, Linf, and L.
- method="AlversonCarney": The "Alverson and Carney (1975) equation" as given in equation 10 of Zhang and Megrey (2006). Requires tmax and K.
- method="Charnov": The "Charnov et al. (2013) equation" as given in the second column of pge 545 of Kenchington (2014). Requires K, Linf, and L.

• method="ZhangMegreyD", method="ZhangMegreyP": The "Zhang and Megrey (2006) equation" as given in their equation 8 but modified for demersal or pelagic fish. Thus, the user must choose the fish type with Rgroup. Requires tmax, K, t0, t50, and b.

- method="RikhterEfanov1": The "Rikhter and Efanov (1976) equation (#2)" as given in the second column of pge 541 of Kenchington (2014) and in Table 6.4 of Miranda and Bettoli (2007). Requires only t50.
- method="RikhterEfanov2": The "Rikhter and Efanov (1976) equation (#1)" as given in the first column of pge 541 of Kenchington (2014). Requires t50, K, t0, and b.

Value

Mmethods returns a charachter vector with a list of methods. If only one method is chosen then metaM returns a single numeric if justM=TRUE or, otherwise, a metaM object that is a list with the following items:

- method: The name for the method within the function (as given in method).
- name: A more descriptive name for the method.
- givens: A vector of values required by the method to estimate M.
- M: The estimated natural mortality rate.

If multiple methods are chosen then a data frame is returned with the method name abbreviation in the method variable and the associated estimated of M in the M variable.

Testing

Kenchington (2014) provided life history parameters for several stocks and used many models to estimate M. I checked the calculations for the PaulyL, PaulyW, HoenigO for Hgroup="all" and Hgroup="fish", HoenigO2 for Hgroup="all" and Hgroup="fish", "JensenK1", "Gislason", "AlversonCarney", "Charnov", "ZhangMegrey", "RikhterEfanov1", and "RikhterEfanov2" methods for three stocks. All results perfectly matched Kenchington's results for Chesapeake Bay Anchovy and Rio Formosa Seahorse. For the Norwegian Fjord Lanternfish, all results perfectly matched Kenchington's results except for when Hgroup="fish" for both HoenigO and HoenigO2.

Results for the Rio Formosa Seahorse data were also tested against results from M. empirical from **fishmethods** for the PaulyL, PaulyW, HoenigO for Hgroup="all" and Hgroup="fish", "Gislason", and "AlversonCarney" methods (the only methods in common between the two packages). All results matched perfectly.

IFAR Chapter

11-Mortality.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Alverson, D.L. and M.J. Carney. 1975. A graphic review of the growth and decay of population cohorts. Journal du Conseil International pour l'Exploration de la Mer. 36:133-143.

Charnov, E.L., H. Gislason, and J.G. Pope. 2013. Evolutionary assembly rules for fish life histories. Fish and Fisheries. 14:213-224.

Gislason, H., N. Daan, J.C. Rice, and J.G. Pope. 2010. Size, growth, temperature and the natural mortality of marine fish. Fish and Fisheries 11:149-158.

Hewitt, D.A. and J.M. Hoenig. 2005. Comparison of two approaches for estimating natural mortality based on longevity. Fishery Bulletin. 103:433-437.

Hoenig, J.M. 1983. Empirical use of longevity data to estimate mortality rates. Fishery Bulletin. 82:898-903.

Jensen, A.L. 1996. Beverton and Holt life history invariants result from optimal trade-off of reproduction and survival. Canadian Journal of Fisheries and Aquatic Sciences. 53:820-822.

Jensen, A.L. 2001. Comparison of theoretical derivations, simple linear regressions, multiple linear regression and principal components for analysis of fish mortality, growth and environmental temperature data. Environmetrics. 12:591-598.

Kenchington, T.J. 2014. Natural mortality estimators for information-limited fisheries. Fish and Fisheries. 14:533-562.

Pauly, D. 1980. On the interrelationships between natural mortality, growth parameters, and mean environmental temperature in 175 fish stocks. Journal du Conseil International pour l'Exploration de la Mer. 39:175-192.

Rikhter, V.A., and V.N. Efanov. 1976. On one of the approaches for estimating natural mortality in fish populations (in Russian). ICNAF Research Document 76/IV/8, 12pp.

Then, A.Y., J.M. Hoenig, N.G. Hall, and D.A. Hewitt. 2015. Evaluating the predictive performance of empirical estimators of natural mortality rate using informatno on over 200 fish species. ICES Journal of Marine Science. 72:82-92.

Zhang, C-I and B.A. Megrey. 2006. A revised Alverson and Carney model for estimating the instantaneous rate of natural mortality. Transactions of the American Fisheries Socity. 135-620-633.

See Also

See M. empirical in **fishmethods** for similar functionality.

```
## List names for available methods
Mmethods()
Mmethods("tmax")

## Simple Examples
metaM("tmax",tmax=20)
metaM("tmax",tmax=20,justM=FALSE)
```

```
## Example Patagonian Sprat ... from Table 2 in Cerna et al. (2014)
## http://www.scielo.cl/pdf/lajar/v42n3/art15.pdf
T <- 11
Linf <- 17.71
K < - 0.78
t0 <- -0.46
tmax <- t0+3/K
t50 <- t0-(1/K)*log(1-13.5/Linf)
metaM("RikhterEfanov1", t50=t50)
metaM("PaulyL",K=K,Linf=Linf,T=T)
metaM("PaulyL",K=K,Linf=Linf,T=T,justM=FALSE)
metaM("HoenigO",tmax=tmax)
metaM("HewittHoenig",tmax=tmax)
metaM("AlversonCarney", K=K, tmax=tmax)
## Example of multiple calculations
\verb|metaM(c("RikhterEfanov1","PaulyL","Hoenig0","HewittHoenig","AlversonCarney")|,
     K=K,Linf=Linf,T=T,tmax=tmax,t50=t50)
## Example of multiple methods using Mmethods
# select some methods
metaM(Mmethods()[-c(4,20,22:24,26)],K=K,Linf=Linf,T=T,tmax=tmax,t50=t50)
# select just the Hoenig methods
metaM(Mmethods("Hoenig"),K=K,Linf=Linf,T=T,tmax=tmax,t50=t50)
```

mrClosed

Estimate initial population size for single or multiple census markrecapture data.

Description

Estimates of the initial population size, along with associated confidence intervals, are constructed from single or multiple census mark-recapture data using a variety of methods. For single census data, the initial population size (N) is estimated from the number of marked animals from a first sample (M), number of captured animals in a second sample (n), and the number of recaptured marked animals in the second sample (m) using either the 'naive' Petersen method or Chapman, Ricker, or Bailey modifications of the Petersen method. Single census data can also be separated by group (e.g., size class) to estimate the initial population size by class and for the overall population size. For multiple census data, the initial population size is estimated from the number of captured animals (n), number of recaptured marked animals (m), the number of marked animals that are marked and returned to the population (R), or the number of extant marked animals prior to the sapmle (M) on each of several samples using either the Schnabel (1938) or Schumacher-Eschmeyer (1943) method.

Usage

```
mrClosed(M = NULL, n = NULL, m = NULL, R = NULL,
method = c("Petersen", "Chapman", "Ricker", "Bailey", "Schnabel",
    "SchumacherEschmeyer"), labels = NULL, chapman.mod = TRUE)
```

```
## S3 method for class 'mrClosed1'
summary(object, digits = 0, incl.SE = FALSE,
  incl.all = TRUE, verbose = FALSE, ...)
## S3 method for class 'mrClosed1'
confint(object, parm = NULL, level = conf.level,
  conf.level = 0.95, digits = 0, type = c("suggested", "binomial",
  "hypergeometric", "normal", "Poisson"), bin.type = c("wilson", "exact",
  "asymptotic"), incl.all = TRUE, verbose = FALSE, ...)
## S3 method for class 'mrClosed2'
summary(object, digits = 0, verbose = FALSE, ...)
## S3 method for class 'mrClosed2'
confint(object, parm = NULL, level = conf.level,
  conf.level = 0.95, digits = 0, type = c("suggested", "normal",
  "Poisson"), verbose = FALSE, ...)
## S3 method for class 'mrClosed2'
plot(x, pch = 19, col.pt = "black",
 xlab = "Marked in Population", ylab = "Prop. Recaptures in Sample",
 loess = FALSE, lty.loess = 2, lwd.loess = 1, col.loess = "gray20",
  trans.loess = 10, span = 0.9, ...)
```

Arguments

m

R

labels

chapman.mod

A numeric representing the number of marked fish from the first sample (single-census), an object from capHistSum() (single- or multiple-census), or numeric vector of marked fish prior to ith samples (multiple-census).

A numeric representing the number of captured fish in the second sample (single-census) or numeric vector of captured fish in ith sample (multiple-census).

A numeric representing the number of recaptured (marked) fish in the second sample (single-census) or numeric vector of recaptured (marked) fish in ith sample (multiple-census).

A numeric vector representing the number of marked fish returned to the population (multiple-census). Note that several references use the number of "new" marks returned to the population rather than the "total" number of marks returned to the population that is used here.

method A single string that identifies the type of calculation method to use in the main function.

A character or character vector used to label the rows of the resulting output matrix when using a single census method separated by groups. Must be the same length as M, n, and m. Defaults to upper-case letters if no values are given.

A logical that represents whether the Chapman modification should be used (=TRUE, default) or not (=FALSE) when performing the Schnabel multiple census method.

object,x	An mrClosed1 or mrClosed2 object.	
digits	The number of decimal digits to round the population estimates to. If incl. SE=TRUE then SE will be rounded to one more decimal place then given in digits.	
incl.SE	A logical that indicates whether the results should include the calculated SE value. See details.	
incl.all	A logical that indicates whether an overall population estimate should be computed when using a single census method that has been separated into subgroups. See details.	
verbose	A logical that indicates whether a reminder of the inputted values and what type of method was used should be printed with the summary and confidence interval results.	
parm	Not used here (included in confint generic).	
level	Same as conf.level but used for compatability with confint generic.	
conf.level	A numeric representing the level of confidence to use for confidence intervals.	
type	A single string that identifies the distribution to use when constructing confidence intervals in confint. See details.	
bin.type	A string that identifies the method used to construct binomial confidence intervals (default is "wilson"). This is only used if type="binomial" in confint. See details of binCI	
pch	A numeric used to indicate the type of plotting character.	
col.pt	a string used to indicate the color of the plotted points.	
xlab	A label for the x-axis.	
ylab	A label for the y-axis.	
loess	A logical that indicates if a loess smoother line (and approximate 95% confidence band) is fit to and shown on plot.	
lty.loess	A single numeric used to indicate the type of line used for the loess line.	
lwd.loess	A single numeric used to indicate the line width of the loess line.	
col.loess	A single string used to indicate the color of the loess line.	
trans.loess	A single numeric that indicates how transparent the loess band should be (larger numbers are more transparent).	
span	A single numeric that controls the degree of smoothing. Values closer to 1 are more smooth.	
	Additional arguments for methods.	

Details

For single census data, the following methods can be used:

- method="Petersen". The 'naive' Petersen as computed using equation 2.1 from Krebs (1989).
- method="Chapman". The Chapman (1951) modification of the Petersen method as computed using equation 2.2 from Krebs (1989).

method="Ricker". The Ricker (1975) modification of the Petersen as computed using equation 3.7 from Ricker (1975). This is basically the same method="Chapman" except that Ricker (1975) did NOT subtract a 1 from the answer in the final step. Thus, the estimate from method="Chapman" will always be one less than the estimate from method="Ricker".

• method="Bailey". The Bailey (1951, 1952) modification of the Petersen as computed using equation 2.3 from Krebs (1989).

If M contains an object from capHistSum and one of Petersen, Chapman, Ricker, or Bailey methods has been selected with method= then n= and m= can be left missing or will be ignored and the needed data will be extracted from the sum portion of the CapHist class object. If the data were not summarized with capHistSum then all of M=, n=, and m= must be supplied by the user.

The population estimate (as computed with the formulas noted in the table above) is extracted with summary. In addition, the standard error of the population estimate (SE) can be extracted by including incl.SE=TRUE. The SE is from equation 3.6 (p. 78) in Ricker (1975) for the Petersen method, from p. 60 (near bottom) of Seber (2002) for the Chapman method, from p. 61 (middle) of Seber (2002) (and as noted on p. 79 of Ricker (1975)) for the Bailey method, and from equation 3.8 (p. 78) in Ricker (1975) for the Ricker method.

Confidence intervals for the initial population size from the single census methods can be constructed using four different distributions as chosen with type= in confint. If type="suggested" then the type of confidence interval suggested by the rules on p. 18 in Krebs (1989) are used. The general methods for constructing confidence intervals for N are described below

- type="hypergeometric". Uses hyperCI. This is experimental at this point.
- type="binomial". Use binCI to construct a confidence interval for m/n (Petersen method) or (m+1)/(n+1) (Chapman, Bailey, Ricker methods), divides M or (M+1) by the CI endpoints, and substract 1 (for the Chapman method).
- type="Poisson". Use poiCI to construct a confidence interval for m (Petersent method) or (m+1) (Chapman, Bailey, Ricker methods), substitute the CI endpoints into the appropriate equation for estimating N, and subtract 1 (for the Chapman method).
- type="normal". Used equation 2.4 (p.20) from Krebs (2002) for the Petersen method. For the other methods, used N+/- Z(0.975)*SE, where the SE was computed as noted above.

If incl.all=TRUE in summary and population estimates have been constructed for multiple sub-groups then an overall population estimate is included by summing the population estimates for the multiple sub-groups. If incl.SE=TRUE, then an overall SE is computed by taking the square root of the summed VARIANCES for the multiple sub-groups.

For multiple census data, the following methods can be declared for use with the method= argument:

- method="Schnabel". The Schnabel (1938) method as computed with equation 3.15 from Ricker (1975).
- method="SchumacherEschmeyer". The Schumacher and Eschmeyer (1943) method as computed with equation 3.12 from Ricker (1975) eqn 3.12.

If M contains an object from capHistSum and the Schnabel or Schumacher-Eschmeyer methods has been chosen then n, m and R can be left missing or will be ignored. In this case, the needed data is extracted from the sum portion of the CapHist class object. Otherwise, the user must supply vectors of results in n, m, and R or M.

The population estimate for each method is extracted with summary. Standard errors for the population estimate can NOT be computed for the Schnabel or Schumacher-Eschmeyer methods (a warning will be produced if incl.SE=TRUE is used).

Confidence intervals for the initial population size using multiple census methods can be constructed using the normal or Poisson distributions for the Schnabel method or the normal distribution for the Shumacher-Eschmeyer method as chosen with type=. If type="suggested" then the type of confidence interval suggested by the rule on p. 32 of Krebs (1989) is used (for the Schnabel method). If type="Poisson" for the Schnabel method then a confidence interval for the sum of m is computed with poiCI and the end points are substituted into the Schnabel equation to produce a CI for the population size. If type="normal" for the Schnabel method then the standard error for the *inverse* of the population estimate is computed as the square root of equation 2.11 from Krebs (1989) or equation 3.16 from Ricker (1975). The standard error for the Schumacher-Eschmeyer method is for the inverse of the population estimate and is computed with equation 2.14 from Krebs (1989) [Note that the divisor in Krebs (1989) is different than the divisor in equation 3.12 in Ricker (1975), but is consistent with equation 4.17 in Seber (2002).] The confidence interval for the *inverse* population estimate is constructed from the inverse population estimate plus/minus at critical value times the standard error for the inverse population estimate. The t critical value uses the number of samples minus 1 for the Schnabel method and the number of samples minus 2 when for the Schumacher-Eschmeyer method according to p. 32 of Krebs (1989) (note that this is different than what Ricker (1975) does). Finally, the confidence interval for the population estimate is obtained by inverting the confidence interval for the inverse population estimate. Note that confidence intervals for the population size when type="normal" may contain negative values (for the upper value) when the population estimate is relatively large and the number of samples is small (say, three) because the intervals are originally constructed on the inverted population estimate and they use the t-distribution.

The plot can be used to identify assumption violations in the Schnabel and Schumacher-Eschmeyer methods (an error will be returned if used with any of the other methods). If the assumptions ARE met then the plot of the proportion of marked fish in a sample versus the cumulative number of marked fish should look linear. A loess line (with approximate 95% confidence bands) can be added to aid interpretation with loess=TRUE. Note, however, that adding the loess line may return a number of warning or produce a non-informative if the number of samples is small (<8).

Value

A list with the following items

- M The number of marked fish from the first sample that was provided.
- n The number of captured fish in the second sample that was provided.
- m The number of recaptured (marked) fish in the second sample that was provided.
- M1 The adjusted (depending on type) number of marked fish from the first sample.
- n1 The adjusted (depending on type) number of captured fish in the second sample.
- m1 The adjusted (depending on type) number of recaptured (marked) fish in the second sample.
- cf A correction factor for the population estimate that depends on type.
- method The type of method used (provided by the user).
- methodLbl A label for the type of method used.

- N The estimated initial population size.
- labels Labels for the rows of summary matrix.

Testing

The results from the single census methods have had the following checks. The population estimates for all methods match reputable sources. The SE for the Chapman and Bailey methods match the results from mrN.single in **fishmethods**, The CI for the Petersen, Chapman, and Bailey methods partially match (are within 1

The results for the multiple census methods have had the following checks. The population estimates for both methods match reputable sources. The intermediate calculations for both methods match those in Krebs (1989). The confidence interval for the Schnabel method using the Poisson distribution does NOT match Krebs (1989). This appears to be a difference in the use poiCI here versus distributional tables in Krebs (i.e., the difference appears to be completely in the critical values from the Poisson distribution). The confidence interval for the Schnabel method using the normal or the Poisson distribution do NOT match Ricker (1975), but there is not enough information in Ricker to determine why (it is likely due to numerical differences on the inverse scale). The confidence interval for the Schumacher-Eschmeyer method do match Krebs (1989) but not Ricker (1975). The Ricker result may be due to different df as noted above.

IFAR Chapter

9-Abundance from Capture-Recapture Data.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Krebs, C.J. 1989. Ecological Methodology. Addison-Welsey Educational Publishing.

Ricker, W.E. 1975. Computation and interpretation of biological statistics of fish populations. Technical Report Bulletin 191, Bulletin of the Fisheries Research Board of Canada.

Seber, G.A.F. 2002. The Estimation of Animal Abundance and Related Parameters. Edward Arnold, second edition.

Schnabel, Z.E. 1938. The estimation of the total fish population of a lake. American Mathematician Monthly, 45:348-352.

Schumacher, F.X. and R.W. Eschmeyer. 1943. The estimation of fish populations in lakes and ponds. Journal of the Tennessee Academy of Sciences, 18:228-249.

See Also

See capHistSum for generating input data from capture histories. See poiCI, binCI, and hyperCI for specifics on functions used in confidence interval constructuion. See mrOpen for handling mark-recapture data in an open population. See SunfishIN in FSAdata for an example to test matching of results with Ricker (1975)' See mrN. single and schnabel in fishmethods for similar functionality.

```
### Single census with no sub-groups
## Petersen estimate -- the default
mr1 <- mrClosed(346,184,49)
summary(mr1)
summary(mr1,verbose=TRUE)
summary(mr1,incl.SE=TRUE)
summary(mr1,incl.SE=TRUE,digits=1)
confint(mr1)
confint(mr1,verbose=TRUE)
confint(mr1,type="hypergeometric")
## Chapman modification of the Petersen estimate
mr2 <- mrClosed(346,184,49,method="Chapman")</pre>
summary(mr2,incl.SE=TRUE)
summary(mr2,incl.SE=TRUE,verbose=TRUE)
### Single census, using capHistSum() results
## data in capture history format
data(BluegillJL)
str(BluegillJL)
ch1 <- capHistSum(BluegillJL)</pre>
mr3 <- mrClosed(ch1)</pre>
summary(mr3,verbose=TRUE)
confint(mr3,verbose=TRUE)
### Single census with sub-groups
marked <-c(93,35,72,16,46,20)
captured <- c(103,30,73,17,39,18)
recaps <- c(20,23,52,15,35,16)
lbls <- c("YOY", "Juvenile", "Stock", "Quality", "Preferred", "Memorable")</pre>
mr4 <- mrClosed(marked,captured,recaps,method="Ricker",labels=lbls)</pre>
summary(mr4)
summary(mr4,incl.SE=TRUE)
summary(mr4,incl.SE=TRUE,verbose=TRUE)
summary(mr4,incl.SE=TRUE,incl.all=FALSE,verbose=TRUE)
confint(mr4)
confint(mr4,verbose=TRUE)
confint(mr4,incl.all=FALSE,verbose=TRUE)
### Multiple Census
## Data in summarized form
data(PikeNY)
## Schnabel method
mr5 <- with(PikeNY,mrClosed(n=n,m=m,R=R,method="Schnabel"))</pre>
plot(mr5,loess=TRUE)
summary(mr5)
summary(mr5,verbose=TRUE)
confint(mr5)
confint(mr5,verbose=TRUE)
```

oddeven 129

```
## Schumacher-Eschmeyer method
mr6 <- with(PikeNY,mrClosed(n=n,m=m,R=R,method="Schumacher"))
summary(mr6)
confint(mr6)

### Capture history data summarized by capHistSum()
data(PikeNYPartial1)
# ignore first column of ID numbers
ch2 <- capHistSum(PikeNYPartial1,cols2ignore="id")

## Schnabel method
mr7 <- mrClosed(ch2,method="Schnabel")
plot(mr7)
summary(mr7)
confint(mr7)</pre>
```

oddeven

Determine if a number is odd or even.

Description

Determine if a number is odd or even.

Usage

```
is.odd(x)
is.even(x)
```

Arguments

Χ

A numeric vector.

Value

A logical vector of the same length as x.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

```
## Individual values
is.odd(1)
is.odd(2)
is.even(3)
is.even(4)
```

perc perc

```
## Vector of values
d <- 1:8
data.frame(d,odd=is.odd(d),even=is.even(d))</pre>
```

perc

Computes the percentage of values in a vector less than or greater than (and equal to) some value.

Description

Computes the percentage of values in a vector less than or greater than (and equal to) a user-supplied value.

Usage

```
perc(x, val, dir = c("geq", "gt", "leq", "lt"), na.rm = TRUE,
  digits = getOption("digits"))
```

Arguments

X	A numeric vector.
val	A single numeric value.
dir	A string that indicates whether the percentage is for values in x that are "greater than and equal" "geq", "greater than" "gt", "less than and equal" "leq", "less than" "lt" the value in val.
na.rm	A logical that indicates whether NA values should be removed (DEFAULT) from x or not.
digits	A single numeric that indicates the number of decimals the percentage should be rounded to.

Details

This function is most useful when used with an apply-type of function.

Value

A single numeric that is the percentage of values in x that meet the criterion in dir relative to val.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

PikeNY 131

Examples

```
## vector of values
( tmp <- c(1:8,NA,NA) )

## percentages excluding NA values
perc(tmp,5)
perc(tmp,5,"gt")
perc(tmp,5,"leq")
perc(tmp,5,"lt")

## percentages including NA values
perc(tmp,5,na.rm=FALSE)
perc(tmp,5,"gt",na.rm=FALSE)
perc(tmp,5,"leq",na.rm=FALSE)
perc(tmp,5,"leq",na.rm=FALSE)</pre>
```

PikeNY

Summarized multiple mark-recapture data for all Northern Pike from Buckthorn Marsh.

Description

Summary results of capture histories (number captured, number of recaptured fish, and number of unmarked fish that were marked) for all Buckthorn Marsh Northern Pike (*Esox lucius*).

Format

A data frame with 21 observations on the following 4 variables:

date Capture date.

- **n** Total fish captured in each sample.
- m Marked fish captured in each sample.
- **R** Marked fish returned to the population.

Topic(s)

- Population size
- Abundance
- · Mark-recapture
- · Schnabel method
- · Schumacher-Eschmeyer method

Source

New York Power Authority. 2004. Use of buckhorn marsh and grand island tributaries by northern pike for spawning and as a nursery. Technical report, New York Power Authority, January 2004. Niagara Power Project (FERC No. 2216).

132 PikeNYPartial1

Examples

data(PikeNY)
str(PikeNY)
head(PikeNY)

PikeNYPartial1

Capture histories (4 samples), in capture history format, of a subset of Northern Pike from Buckthorn Marsh.

Description

Each line consists of the capture history over four samples of Northern Pike (*Esox lucius*) in Buckthorn Marsh. This file contains the capture histories for only those pike captured from April 1-4.

Format

A data frame with 57 observations on the following 4 variables.

id A unique identification numbers.

first Indicator variable for the first sample (1=captured).

second Indicator variable for the second sample (1=captured).

third Indicator variable for the third sample (1=captured).

fourth Indicator variable for the fourth sample (1=captured).

Topic(s)

- · Population size
- Abundance
- · Mark-recapture
- · Schnabel method
- Schumacher-Eschmeyer method
- · Capture history

Source

New York Power Authority. 2004. Use of buckhorn marsh and grand island tributaries by northern pike for spawning and as a nursery. Technical report, New York Power Authority, January 2004. Niagara Power Project (FERC No. 2216).

```
data(PikeNYPartial1)
str(PikeNYPartial1)
head(PikeNYPartial1)
```

plotBinResp 133

variable.	plotBinResp	Plots a binary response variable versus a quantitative explanatory variable.
-----------	-------------	--

Description

A function to plot a binary response variable versus a quantitative explanatory variable.

Usage

```
plotBinResp(x, ...)

## Default S3 method:
plotBinResp(x, y, xlab = paste(deparse(substitute(x))),
  ylab = paste(deparse(substitute(y))), plot.pts = TRUE, col.pt = "black",
  transparency = NULL, plot.p = TRUE, breaks = 25, p.col = "blue",
  p.pch = 3, p.cex = 1.25, yaxis1.ticks = seq(0, 1, 0.1),
  yaxis1.lbls = c(0, 0.5, 1), yaxis2.show = TRUE, ...)

## S3 method for class 'formula'
plotBinResp(x, data = NULL, xlab = names(mf)[2],
  ylab = names(mf)[1], ...)
```

Arguments

х	A quantitative explanatory variable or a formula of the form factor~quant.
У	A binary response variable.
xlab	A string for labelling the x-axis.
ylab	A string for labelling the y-axis.
plot.pts	A logical that indicates (TRUE (default)) whether the points should be plotted (TRUE; default) or not (FALSE).
col.pt	A string used to indicate the color of the plotted points. Will be transparent unless transparency=1.
transparency	A numeric that indicates how many points would be plotted on top of each other before the 'point' would have the full col.pt color. The reciprocal of this value is the alpha transparency value.
plot.p	A logical that indicates if the proportion for categorized values of \boldsymbol{X} are plotted (TRUE; default).
breaks	A number that indicates how many intervals over which to compute proportions or a numeric vector that contains the endpoints of the intervals over which to compute proportions if plot.p=TRUE.
p.col	A color to plot the proportions.
p.pch	A plotting character for plotting the proportions.
p.cex	A character expansion factor for plotting the proportions.

134 plotBinResp

yaxis1.ticks	A numeric vector that indicates where tick marks should be placed on the left y-axis (for the proportion of 'successes').
yaxis1.lbls	A numeric vector that indicates labels for the tick marks on the left y-axis (for the proportion of 'successes').
yaxis2.show	A logical that indicates whether the right y-axis should be created (=TRUE; default) or not.
data	The data frame from which the formula should be evaluated.
	Other arguments to be passed to the plot functions.

Details

This function produces a plot that can be used to visualize the density of points for a binary response variable as a function of a quantitative explanatory variable. In addition, the proportion of "1"s for the response variable at various "levels" of the explanatory variable are shown.

Value

None. However, a plot is produced.

Note

This function is meant to allow newbie students the ability to visualize the data corresponding to a binary logistic regression without getting "bogged-down" in the gritty details of how to produce this plot.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

See Also

fitPlot and cdplot.

```
## NASA space shuttle o-ring failures -- from graphics package
fail <- factor(c(2,2,2,2,1,1,1,1,1,1,2,1,2,1,1,1,1,2,1,1,1,1),
levels = 1:2, labels = c("no","yes"))
temperature <- c(53,57,58,63,66,67,67,67,68,69,70,70,70,70,72,73,75,75,76,76,78,79,81)
d <- data.frame(fail,temperature)

## Default plot (using formula notation)
plotBinResp(fail~temperature,data=d)

## Controlling where proportions are computed with a sequence in breaks
plotBinResp(fail~temperature,data=d,breaks=seq(50,85,5))

## Controlling where proportions are computed with an integer in breaks
plotBinResp(fail~temperature,data=d,breaks=10)</pre>
```

poiCI 135

```
## Controlling where proportions are computed at each value of x
plotBinResp(fail~temperature,data=d,breaks=NULL)

## Don't plot points, just plot proportions
plotBinResp(fail~temperature,data=d,plot.pts=FALSE)

## Don't plot proportions, just plot points
plotBinResp(fail~temperature,data=d,plot.p=FALSE)

## Change points colors, and eliminate transparency
plotBinResp(fail~temperature,data=d,col.pt="red",transparency=1)

## Remove the right y-axis
plotBinResp(fail~temperature,data=d,yaxis2.show=FALSE)

## Change left y-axis ticks
plotBinResp(fail~temperature,data=d,yaxis1.ticks=c(0,1),yaxis1.lbls=c(0,1))
```

poiCI

Confidence interval for Poisson rate parameter.

Description

Computes a confidence interval for the Poisson mean rate parameter.

Usage

```
poiCI(x, conf.level = 0.95)
```

Arguments

x A number representing the number of observed successes.

conf.level A number that indicates the level of confidence to use for constructing confidence intervals (default is 0.95).

Details

Computes a CI for the Poisson mean using the method described in Ulm (1990), though this method was earlier described by Liddell (1984) and possibly Garwood (1936) as noted in van der Gulden and Verbeck (1992). Thank you to Jerry Lewis for clarifications to the historical citations of this method.

Value

A 1x2 matrix that contains the lower and upper confidence interval bounds.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

psdAdd psdAdd

References

Garwood, F. 1936. Fiducial limits for the Poisson distribution. Biometrika. 28(3/4):437-442.

Liddell, F.D. 1984. Simple exact analysis of the standardised mortality ratio. Journal of Epidemiology and Community Health. 38(1):85-88.

Ulm, K. 1990. A simple method to calculate the confidence interval of a standardized mortality ratio. American Journal of Epidemiology 131(2):373-375.

vand der Gulden, J.W.J. and A.L.M. Verbeck. 1992. Re: "A simple method to calculate the confidence interval of a standardized mortality ratio (SMR)". American Journal of Epidemiology 136(9):1170-1171.

See Also

See pois. exact in **epitools** for similar functionality.

Examples

```
poiCI(12)
```

psdAdd

Create vector of Gabelhouse lengths for each species from an entire data frame.

Description

Creates a vector of the Gabelhouse lengths specific to a species for all individuals in an entire data frame.

Usage

```
psdAdd(len, ...)
## Default S3 method:
psdAdd(len, spec, units = c("mm", "cm", "in"),
   use.names = TRUE, addSpec = NULL, addLens = NULL, verbose = TRUE, ...)
## S3 method for class 'formula'
psdAdd(len, data = NULL, units = c("mm", "cm", "in"),
   use.names = TRUE, addSpec = NULL, addLens = NULL, verbose = TRUE, ...)
```

Arguments

len

A numeric vector that contains lengths measurements or a formula of the form len~spec where "len" generically represents the length variable and "spec" generically represents the species variable. Note that this formula can only contain two variables and must have the length variable on the left-hand-side and the species variable on the right-hand-side.

psdAdd 137

spec	A character or factor vector that contains the specicas names. Not used if len is a formula.
units	A string that indicates the type of units used for the lengths. Choices are mm for millimeters (DEFAULT), cm for centimeters, and in for inches.
use.names	A logical that indicates whether the vector returned is numeric (=FALSE) or word (=TRUE; default) representations of the Gabelhouse lengths. See details.
addSpec	A character vector of species names for whiach addLens will be provided.
addLens	A numeric vector of lengths that should be used for the species in addSpec in addition to the Gabelhouse lengths. See examples.
verbose	A logical that indicates whether detailed messages about species without Gabelhouse lengths or with no recorded values should be printed or not.
data	A data.frame that minimally contains the length measurements and species names if 1en is a formula.
	Not used.

Details

This computes a vector that contains the Gabelhouse lengths specific to each species for all individuals in an entire data frame. The vector can be appended to an existing data.frame to create a variable that contains the Gabelhous lengths for each individual. The Gabelhouse length value will be NA for each individual for which a Gabelhouse length definitision do not exist in PSD1it. Individuals shorter than "stock" length will be listed as zero if use.names=TRUE or 0 if use.names=FALSE. See the examples for one method for changing species names to something that this function will recognize.

Additional lengths to be use for each species can be included by giving a vector of species names in addSpec and a corresponding vector of additional lengths in addLens. Note, however, that use.names will be reset to FALSE if addSpec and addLens are specified as there is no way to order the names when additional lengths are used.

Value

A numeric or factor vector that contains the Gabelhouse length categories.

IFAR Chapter

6-Size Structure.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL. Guy, C.S., R.M. Neumann, and D.W. Willis. 2006. New terminology for proportional stock density (PSD) and relative stock density (RSD): proportional size structure (PSS). Fisheries 31:86-87.

psdAdd psdAdd

Guy, C.S., R.M. Neumann, D.W. Willis, and R.O. Anderson. 2006. Proportional size distribution (PSD): A further refinement of population size structure index terminology. Fisheries 32:348. Willis, D.W., B.R. Murphy, and C.S. Guy. 1993. Stock density indices: development, use, and

limitations. Reviews in Fisheries Science 1:203-222.

See Also

psdVal, psdCalc, psdPlot, PSDlit, and wrAdd for related functions. See mapvalues for help in changing species names to match those in PSDlit.

```
## Create random data for three species
# only for repeatability
set.seed(345234534)
dbg <- data.frame(species=factor(rep(c("Bluegill"),30)),tl=round(rnorm(30,130,50),0))</pre>
dbg$wt <- round(4.23e-06*dbg$tl^3.316+rnorm(30,0,10),1)
dlb <- data.frame(species=factor(rep(c("Largemouth Bass"),30)),tl=round(rnorm(30,350,60),0))</pre>
dlb$wt <- round(2.96e-06*dlb$tl^3.273+rnorm(30,0,60),1)
dbt <- data.frame(species=factor(rep(c("Bluefin Tuna"),30)),tl=round(rnorm(30,1900,300),0))</pre>
dbt$wt <- round(4.5e-05*dbt$tl^2.8+rnorm(30,0,6000),1)
df <- rbind(dbg,dlb,dbt)</pre>
str(df)
## Examples (non-dplyr)
# Add variable using category names -- formula notation
df$PSD <- psdAdd(tl~species,data=df)</pre>
head(df)
# Add variable using category names -- non-formula notation
df$PSD1 <- psdAdd(df$tl,df$species)</pre>
head(df)
# Add variable using length values as names
df$PSD2 <- psdAdd(tl~species,data=df,use.names=FALSE)</pre>
# Add additional length and name for Bluegill
df$PSD3 <- psdAdd(tl~species,data=df,addSpec="Bluegill",addLens=175)</pre>
head(df)
# Add additional lengths and names for Bluegill and Largemouth Bass from a data.frame
addls <- data.frame(species=c("Bluegill","Largemouth Bass","Largemouth Bass"),</pre>
                    lens=c(175,254,356))
df$psd4 <- psdAdd(t1~species,data=df,addSpec=addls$species,addLens=addls$lens)</pre>
head(df)
## All of the above but using dplyr
if (require(dplyr)) {
 df <- df %>%
    mutate(PSD1A=psdAdd(tl,species)) %>%
   mutate(PSD2A=psdAdd(t1,species,use.names=FALSE)) %>%
    mutate(psd3a=psdAdd(t1,species,addSpec="Bluegill",addLens=175)) %>%
    mutate(psd4a=psdAdd(t1,species,addSpec=addls$species,addLens=addls$lens))
}
df
```

psdCalc 139

nad	\sim	1 ~
psd	Lа	T.C.

Convenience function for calculating PSD-X and PSD X-Y values.

Description

Convenience function for calculating (traditional) PSD-X and (incremental) PSD X-Y values for all Gabelhouse lengths and increments thereof.

Usage

```
psdCalc(formula, data, species, units = c("mm", "cm", "in"),
  method = c("multinomial", "binomial"), conf.level = 0.95,
  addLens = NULL, addNames = NULL, justAdds = FALSE, what = c("all",
  "traditional", "incremental", "none"), drop0Est = TRUE,
  showIntermediate = FALSE, digits = 0)
```

Arguments

formula	A formula of the form ~length where "length" generically represents a variable in data that contains the observed lengths. Note that this formula may only contain one variable and it must be numeric.	
data	A data frame that minimally contains the observed lengths given in the variable in formula.	
species	A string that contains the species name for which Gabelhouse lengths exist. See psdVal for details.	
units	A string that indicates the type of units used for the lengths. Choices are mm for millimeters (DEFAULT), cm for centimeters, and in for inches.	
method	A character that identifies the confidence interval method to use. See details in psdCI.	
conf.level	A number that indicates the level of confidence to use for constructing confidence intervals (default is 0.95).	
addLens	A numeric vector that contains minimum lengths for additional categories. See psdVal for details.	
addNames	A string vector that contains names for the additional lengths added with addLens. See psdVal for details.	
justAdds	A logical that indicates whether just the values related to the length sin addLens should be returned.	
what	A string that indicates the type of PSD values that will be printed. See details.	
drop0Est	A logical that indicates whether the PSD values that are zero should be dropped from the output.	

showIntermediate

A logical that indicates whether the number of fish in the category and the number of stock fish (i.e., "intermediate" values) should be included in the returned matrix. Default is to not include these values.

psdCalc psdCalc

digits

A numeric that indicates the number of decimals to round the result to. Default is zero digits following the recommendation of Neumann and Allen (2007).

Details

Computes the (traditional) PSD-X and (incremental) PSD X-Y values, with associated confidence intervals, for each Gabelhouse length. All PSD-X and PSD X-Y values are printed if what="all" (DEFAULT), only PSD-X values are printed if what="traditional", only PSD X-Y values are printed if what="incremental", and nothing is printed (but the matrix is still returned) if what="none".

Confidence intervals can be computed with either the multinomial (Default) or binomial distribution as set in method. See details in psdCI for more information.

Value

A matrix with columns that contain the computed PSD-X or PSD X-Y values and associated confidence intervals. If showIntermediate=TRUE then the number of fish in the category and the number of stock fish will also be shown.

Testing

Point estimate calculations match those constructed "by hand."

IFAR Chapter

6-Size Structure.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Guy, C.S., R.M. Neumann, and D.W. Willis. 2006. New terminology for proportional stock density (PSD) and relative stock density (RSD): proportional size structure (PSS). Fisheries 31:86-87.

Guy, C.S., R.M. Neumann, D.W. Willis, and R.O. Anderson. 2006. Proportional size distribution (PSD): A further refinement of population size structure index terminology. Fisheries 32:348.

Neumann, R. M. and Allen, M. S. 2007. Size structure. In Guy, C. S. and Brown, M. L., editors, Analysis and Interpretation of Freshwater Fisheries Data, Chapter 9, pages 375-421. American Fisheries Society, Bethesda, MD.

Willis, D.W., B.R. Murphy, and C.S. Guy. 1993. Stock density indices: development, use, and limitations. Reviews in Fisheries Science 1:203-222.

See Also

See psdVal, psdPlot, psdAdd, PSDlit, tictactoe, lencat, and rcumsum for related functionality.

psdCalc 141

```
## Random length data
# suppose this is yellow perch to the nearest mm
yepdf <- data.frame(yepmm=round(c(rnorm(100,mean=125,sd=15),rnorm(50,mean=200,sd=25),</pre>
                    rnorm(20, mean=300, sd=40)), 0),
                    species=rep("Yellow Perch",170))
psdCalc(~yepmm,data=yepdf,species="Yellow perch",digits=1)
psdCalc(~yepmm,data=yepdf,species="Yellow perch",digits=1,drop0Est=TRUE)
## all values above stock value (troubleshooting a problem)
yepdf2 <- subset(yepdf,yepmm>=130)
psdCalc(~yepmm,data=yepdf2,species="Yellow perch")
## all values above quality value (troubleshooting a problem)
vepdf3 <- subset(yepdf,yepmm>=200)
psdCalc(~yepmm,data=yepdf3,species="Yellow perch")
psdCalc(~yepmm,data=yepdf3,species="Yellow perch",drop0Est=TRUE)
## all values below memorable value (troubleshooting a problem)
yepdf4 <- subset(yepdf,yepmm<300)</pre>
psdCalc(~yepmm,data=yepdf4,species="Yellow perch")
## add a length
psdCalc(~yepmm, data=yepdf, species="Yellow perch", addLens=150)
## add lengths with names
psdCalc(~yepmm,data=yepdf,species="Yellow perch",addLens=150,addNames="minLen")
psdCalc(~vepmm,data=yepdf,species="Yellow perch",addLens=c("minLen"=150))
psdCalc(~yepmm,data=yepdf,species="Yellow perch",addLens=c(150,275),addNames=c("minSlot","maxSlot"))
psdCalc(~yepmm,data=yepdf,species="Yellow perch",addLens=c("minLen"=150,"maxslot"=275))
## add lengths with names, return just those values that use those lengths
psdCalc(~yepmm,data=yepdf,species="Yellow perch",addLens=c("minLen"=150),justAdds=TRUE)
psdCalc(~yepmm,data=yepdf,species="Yellow perch",addLens=c("minLen"=150),justAdds=TRUE,
        what="traditional")
psdCalc(~yepmm,data=yepdf,species="Yellow perch",addLens=c(150,275),addNames=c("minSlot","maxSlot"),
        justAdds=TRUE)
psdCalc(~yepmm,data=yepdf,species="Yellow perch",addLens=c(150,275),addNames=c("minSlot","maxSlot"),
        justAdds=TRUE, what="traditional")
## different output types
psdCalc(~yepmm,data=yepdf,species="Yellow perch",addLens=150,what="traditional")
psdCalc(~yepmm,data=yepdf,species="Yellow perch",addLens=150,what="incremental")
psdCalc(~yepmm,data=yepdf,species="Yellow perch",addLens=150,what="none")
## Show intermediate values
psdCalc(~yepmm,data=yepdf,species="Yellow perch",showInterm=TRUE)
psdCalc(~yepmm,data=yepdf,species="Yellow perch",what="traditional",showInterm=TRUE)
psdCalc(~yepmm,data=yepdf,species="Yellow perch",what="incremental",showInterm=TRUE)
## Control the digits
psdCalc(~yepmm,data=yepdf,species="Yellow perch",digits=1)
```

142 psdCI

	~	_
bsa	L.	L

Compute confidence intervals for PSD-X and PSD X-Y values.

Description

Compute confidence intervals for (traditional) PSD-X and (incremental) PSD X-Y values as requested by the user.

Usage

```
psdCI(indvec, ptbl, n, method = c("binomial", "multinomial"),
bin.type = c("wilson", "exact", "asymptotic"), conf.level = 0.95,
label = NULL, digits = 1)
```

Arguments

indvec	A numeric vector of 0s and 1s that identify the linear combination of proprtions from ptbl that the user is interested in. See details.
ptbl	A numeric vector or array that contains the proportion or precentage of all individuals in each length category. See details.
n	A single numeric of the number of fish used to construct ptbl.
method	A string that identifies the confidence interval method to use. See details.
bin.type	A string that identifies the type of method to use for calculation of the confidence intervals when Rmethod="binomial". See details of binCI.
conf.level	A number that indicates the level of confidence to use for constructing confidence intervals (default is 0.95).
label	A single string that can be used to label the row of the output matrix.
digits	A numeric that indicates the number of decimals to round the result to.

Details

Computes confidence intervals for (traditional) PSD-X and (incremental) PSD X-Y values. Two methods can be used as chosen with method=. If method="binomial" then the binomial distribution (via binCI()) is used. If method="multinomial" then the multinomial method described by Brenden et al. (2008) is used. This function is defined to compute one confidence interval so method="binomial" is the default. See examples and psdCalc for computing several simultaneous confidence intervals.

A table of proportions within each length category is given in ptb1. If ptb1 has any values greater than 1 then it is assumed that a table of percentages was supplied and the entire table will be divided by 100 to continue. The proportions must sum to 1 (with some allowance for rounding).

A vector of length equal to the length of ptbl is given in indvec which contains zeroes and ones to identify the linear combination of values in ptbl to use to construct the confidence intervals. For example, if ptbl has four proportions then indvec=c(1,0,0,0) would be used to construct a confidence interval for the population proportion in the first category. Alternatively, indvec=c(0,0,1,1) would be used to construct a confidence interval for the population proportion in the last two categories. This vector must not contain all zeroes or all ones.

psdCI 143

Value

A matrix with columns that contain the computed PSD-X or PSD X-Y value and the associated confidence interval. The confidence interval values were set to zero or 100 if the computed value was negative or greater than 100, respectively.

Testing

The multinomial results match the results given in Brendent et al. (2008).

IFAR Chapter

6-Size Structure.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL. Brenden, T.O., T. Wagner, and B.R. Murphy. 2008. Novel tools for analyzing proportional size distribution index data. North American Journal of Fisheries Management 28:1233-1242.

See Also

See psdVal, psdPlot, psdAdd, PSDlit, tictactoe, lencat, and rcumsum for related functionality.

```
## from Brenden et al. (2008)
ipsd <- c(0.130,0.491,0.253,0.123)
n <- 445
## single binomial
psdCI(c(0,0,1,1),ipsd,n=n)
psdCI(c(1,0,0,0),ipsd,n=n,label="PSD S-Q")
## single multinomial
psdCI(c(0,0,1,1),ipsd,n=n,method="multinomial")
psdCI(c(1,0,0,0),ipsd,n=n,method="multinomial",label="PSD S-Q")
## multiple multinomials (but see psdCalc())
lbls <- c("PSD S-Q","PSD Q-P","PSD P-M","PSD M-T","PSD","PSD-P")</pre>
imat <- matrix(c(1,0,0,0,</pre>
                  0,1,0,0,
                  0,0,1,0,
                  0,0,0,1,
                  0,1,1,1,
                  0,0,1,1),nrow=6,byrow=TRUE)
rownames(imat) <- lbls</pre>
imat
```

144 PSDlit

```
mcis <- t(apply(imat,MARGIN=1,FUN=psdCI,ptbl=ipsd,n=n,method="multinomial"))
colnames(mcis) <- c("Estimate","95% LCI","95% UCI")
mcis

## Multiple "Bonferroni-corrected" (for six comparisons) binomial method
bcis <- t(apply(imat,MARGIN=1,FUN=psdCI,ptbl=ipsd,n=n,conf.level=1-0.05/6))
colnames(bcis) <- c("Estimate","95% LCI","95% UCI")
bcis</pre>
```

PSDlit

Gabelhouse five-cell length categories for various species.

Description

Cutoffs for the Gabelhouse five-cell length categories for a variety of species.

Format

A data frame of 58 observations on the following 11 variables:

```
species Species name.

stock.in Stock length in inches.

quality.in Quality length in inches.

preferred.in Preferred length in inches.

memorable.in Memorable length in inches.

trophy.in Trophy length in inches.

stock.cm Stock length in cm.

quality.cm Quality length in cm.

preferred.cm Preferred length in cm.

memorable.cm Memorable length in cm.

trophy.cm Trophy length in cm.
```

Topic(s)

- Size structure
- Proportional size structure
- · Relative stock density
- · Proportional stock density

IFAR Chapter

6-Size Structure.

psdPlot 145

Source

Original summary table from Dr. Michael Hansen, University of Wisconsin-Stevens Point. Additional species have been added by the package author from the literature.

References

```
Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.
```

See Also

See psdVal, psdCalc, psdPlot, psdAdd, and tictactoe for related functionality.

Examples

```
data(PSDlit)
str(PSDlit)
head(PSDlit)
```

psdPlot

Length-frequency histogram with Gabelhouse lengthshighlighted.

Description

Constructs a length-frequency histogram with Gabelhouse lengths highlighted.

Usage

```
psdPlot(formula, data, species = "List", units = c("mm", "cm", "in"),
  startcat = 0, w = 1, justPSDQ = FALSE, main = "", xlab = "Length",
  ylab = "Number", xlim = NULL, ylim = c(0, max(h$counts)),
  substock.col = "white", stock.col = "gray90", psd.col = "black",
  psd.lty = 2, psd.lwd = 1, show.abbrevs = TRUE, psd.add = TRUE,
  psd.pos = "topleft", psd.cex = 0.75, ...)
```

Arguments

formula	A formula of the form ~length where "length" generically represents a variable in data that contains length measurements. Note that this formula can only contain one variable.
data	A data.frame that minimally contains the length measurements given in the variable in the formula.
species	A string that contains the species name for which five-cell length categories exist. See psdVal for details.
units	A string that indicates the type of units used for the length measurements. Choices are mm for millimeters (DEFAULT), cm for centimeters, and in for inches.
startcat	A number that indicates the beginning of the first length-class.

146 psdPlot

W	A number that indicates the width of length classes to create.
justPSDQ	A logical that indicates whether just stock and quality (for PSD-Q calculations) categories should be used. If FALSE (default) then the five-cell categories of Gabelhouse will be used.
main	A string that serves as the main label for the histogram.
xlab	A string that serves as the label for the x-axis.
ylab	A string that serves as the label for the y-axis.
xlim	A numeric vector of length two that indicates the minimum and maximum values for the x-axis.
ylim	A numeric vector of length two that indicates the minimum and maximum values for the y-axis.
substock.col	A string that indicates the color to use for the bars representing under-stock size fish.
stock.col	A string that indicates the color to use for the bars representing stock size fish.
psd.col	A string that indicates the color to use for the vertical lines at the PSD category values.
psd.lty	A numeric that indicates the line type to use for the vertical lines at the PSD category values.
psd.lwd	A numeric that indicates the line width to use for the vertical lines at the PSD category values.
show.abbrevs	A logical that indicates if the abbreviations for the Gabelhouse length categories should be added to the top of the plot.
psd.add	A logical that indicates if the calculated PSD values should be added to the plot (Default is TRUE).
psd.pos	A string that indicates the position for the legend text.
psd.cex	A numeric value that indicates the character expansion for the legend text.
	Arguments to be passed to the low-level plotting functions.

Details

Constructs a length-frequency histogram with the stock-sized fish highlighted, the Gabelhouse lengths marked by vertical lines, and the (traditional) PSD-X values superimposed.

The length of fish plotted on the x-axis can be controlled with xlim, however, the minimum value in xlim must be less than the stock length for that species.

Value

None. However, a graphic is produced.

IFAR Chapter

6-Size Structure.

psdPlot 147

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Guy, C.S., R.M. Neumann, and D.W. Willis. 2006. New terminology for proportional stock density (PSD) and relative stock density (RSD): proportional size structure (PSS). Fisheries 31:86-87.

Guy, C.S., R.M. Neumann, D.W. Willis, and R.O. Anderson. 2006. Proportional size distribution (PSD): A further refinement of population size structure index terminology. Fisheries 32:348.

Willis, D.W., B.R. Murphy, and C.S. Guy. 1993. Stock density indices: development, use, and limitations. Reviews in Fisheries Science 1:203-222.

See Also

See psdVal, psdCalc, psdAdd, PSDlit, lencat, tictactoe, lencat, and rcumsum for related functionality.

```
## Random length data
# suppose this is yellow perch to the nearest mm
mm <- c(rnorm(100, mean=125, sd=15), rnorm(50, mean=200, sd=25), rnorm(20, mean=300, sd=40))
# same data to the nearest 0.1 cm
cm < - mm/10
# same data to the nearest 0.1 in
inch <- mm/25.4
# put together as data.frame
df <- data.frame(mm,cm,inch)</pre>
## Example graphics
op <- par(mar=c(3,3,2,1),mgp=c(1.7,0.5,0))
# mm data using 10-mm increments
psdPlot(~mm,data=df,species="Yellow perch",w=10)
# cm data using 1-cm increments
psdPlot(~cm, data=df, species="Yellow perch", units="cm", w=1)
# inch data using 1-in increments
psdPlot(~inch,data=df,species="Yellow perch",units="in",w=1)
# same as first with some color changes
psdPlot(~mm,data=df,species="Yellow perch",w=10,substock.col="gray90",stock.col="gray30")
# ... but without the PSD values
psdPlot(~mm,data=df,species="Yellow perch",w=10,psd.add=FALSE)
# ... demonstrate use of xlim
psdPlot(~mm,data=df,species="Yellow perch",w=10,xlim=c(100,300))
## different subsets of fish
# ... without any sub-stock fish
brks <- psdVal("Yellow Perch")</pre>
```

148 psdVal

```
tmp <- Subset(df,mm>brks["stock"])
psdPlot(~mm,data=tmp,species="Yellow perch",w=10)
# ... without any sub-stock or stock fish
tmp <- Subset(df,mm>brks["quality"])
psdPlot(~mm,data=tmp,species="Yellow perch",w=10)
# ... with only sub-stock fish (don't run ... setup to give an error)
tmp <- Subset(df,mm<br/>brks["stock"])
## Not run: psdPlot(~mm,data=tmp,species="Yellow perch",w=10)
```

psdVal

Finds Gabelhouse lengths (for PSD calculations) for a species.

Description

Returns a vector with the five Gabelhouse lengths for a chosen species.

Usage

```
psdVal(species = "List", units = c("mm", "cm", "in"), incl.zero = TRUE,
  addLens = NULL, addNames = NULL)
```

Arguments

species	A string that contains the species name for which to find Gabelhous lengths. See details.
units	A string that indicates the units for the returned lengths. Choices are mm for millimeters (DEFAULT), cm for centimeters, and in for inches.
incl.zero	A logical that indicates if a zero is included in the first position of the returned vector (DEFAULT) or not. This position will be names as "substock". See details.
addLens	A numeric vector that contains minimum length definitions for additional categories. See details.
addNames	A string vector that contains names for the additional length categories added with addLens. See details.

Details

Finds the Gabelhouse lengths from the data(PSDlit) data frame for the species given in species. The species name must be spelled exactly (within capitalization differences) as it appears in data(PSDlit) (type psdVal() to see the list of species and how they are spelled).

A zero is included in the first postion of the returned vector if incl.zero=TRUE. This is useful when computing PSD values with a data frame that contains fish smaller than the stock length.

Additional lengths can be added to the returned vector with addLens. Names for these lengths can be included in addNames. If addNames is non-NULL then it must be of the same length as addLens.

psdVal 149

If addLens is non-NULL but addNames is NULL then the default names will be the same as the lengths in addLens. The addLens argument is useful for calculating PSD values that are different from the Gabelhouse lengths.

Value

A vector of minimum values for length categories for the chosen species.

IFAR Chapter

6-Size Structure.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Guy, C.S., R.M. Neumann, and D.W. Willis. 2006. New terminology for proportional stock density (PSD) and relative stock density (RSD): proportional size structure (PSS). Fisheries 31:86-87.

Guy, C.S., R.M. Neumann, D.W. Willis, and R.O. Anderson. 2006. Proportional size distribution (PSD): A further refinement of population size structure index terminology. Fisheries 32:348.

Willis, D.W., B.R. Murphy, and C.S. Guy. 1993. Stock density indices: development, use, and limitations. Reviews in Fisheries Science 1:203-222.

See Also

See psdCalc, psdPlot, psdAdd, PSDlit, tictactoe, lencat, and rcumsum for related functionality.

```
# List all of the species
psdVal()
# Demonstrate typical usages
psdVal("Yellow perch")
psdVal("Walleye",units="cm")
psdVal("Bluegill",units="in")
psdVal("Bluegill",units="in",incl.zero=FALSE)
psdVal("Bluegill")
# Demonstrate that it will work with mis-capitalization
psdVal("bluegill")
psdVal("Yellow Perch")
# Demonstrate adding in user-defined categories
psdVal("Bluegill",units="in",addLens=7)
psdVal("Bluegill",units="in",addLens=7,addNames="MinLen")
psdVal("Bluegill",units="in",addLens=c(7,9),addNames=c("MinSlot","MaxSlot"))
psdVal("Bluegill",units="in",addLens=c("MinLen"=7))
psdVal("Bluegill",units="in",addLens=c("MinSlot"=7,"MaxSlot"=9))
```

150 rcumsum

rcumsum

Computes the prior to or reverse cumulative sum of a vector.

Description

Computes the prior-to (i.e., the cumulative sum prior to but not including the current value) or the reverse (i.e., the number that large or larger) cumulative sum of a vector. Also works for 1-dimensional tables, matrices, and data.frames, though it is best used with vectors.

Usage

```
rcumsum(x)
pcumsum(x)
```

Arguments

Х

a numeric object.

Value

A numeric vector that contains the prior-to or reverse cumulative sums.

Note

An NA in the vector causes all returned values at and after the first NA for pcumsum and at and before the last NA for rcumsum to be NA. See the examples.

Author(s)

```
Derek H. Ogle, <derek@derekogle.com>
```

See Also

cumsum.

```
rcum=rcumsum(vals))
## Example with NA
vals <- c(1,2,NA,3,NA,4)
cbind(vals,
      cum=cumsum(vals),
      pcum=pcumsum(vals),
      rcum=rcumsum(vals))
## Example with a matrix
mat <- matrix(c(1,2,3,4,5),nrow=1)
cumsum(mat)
pcumsum(mat)
rcumsum(mat)
## Example with a table (must be 1-d)
df <- sample(1:10,100,replace=TRUE)</pre>
tbl <- table(df)
cumsum(tbl)
pcumsum(tbl)
rcumsum(tbl)
## Example with a data.frame (must be 1-d)
df <- sample(1:10,100,replace=TRUE)</pre>
tbl <- as.data.frame(table(df))[,-1]
cumsum(tbl)
pcumsum(tbl)
rcumsum(tbl)
```

removal

Population estimates for k-, 3-, or 2-pass removal data.

Description

Computes estimates, with confidence intervals, of the population size and probability of capture from the number of fish removed in k-, 3-, or 2-passes in a closed population.

Usage

```
removal(catch, method = c("CarleStrub", "Zippin", "Seber3", "Seber2",
   "RobsonRegier2", "Moran", "Schnute"), alpha = 1, beta = 1,
   CS.se = c("Zippin", "alternative"), conf.level = 0.95,
   just.ests = FALSE, Tmult = 3)

## S3 method for class 'removal'
summary(object, parm = c("No", "p", "p1"),
   digits = getOption("digits"), verbose = FALSE, ...)

## S3 method for class 'removal'
```

```
confint(object, parm = c("No", "p"), level = conf.level,
  conf.level = NULL, digits = getOption("digits"), verbose = FALSE, ...)
```

Arguments

catch	A numerical vector of catch at each pass.
method	A single string that identifies the removal method to use. See details.
alpha	A single numeric value for the alpha parameter in the CarleStrub method (default is 1).
beta	A single numeric value for the beta parameter in the CarleStrub method (default is 1).
CS.se	A single string that identifies whether the SE in the CarleStrub method should be computed according to Seber or Zippin.
conf.level	A single number representing the level of confidence to use for constructing confidence intervals. This is sent in the main removal function rather than confint.
just.ests	A logical that indicates whether just the estimates (=TRUE) or the return list (=FALSE; default; see below) is returned.
Tmult	A single numeric that will be multiplied by the total catch in all samples to set the upper value for the range of population sizes when minimizing the log-likelihood and creating confidence intervals for the Moran and Schnute method. Large values are much slower to compute, but too low of value can result in missing the best estimate.
object	An object saved from removal().
parm	A specification of which parameters are to be given confidence intervals, either a vector of numbers or a vector of names. If missing, all parameters are considered.
digits	A single numeric that controls the number of decimals in the output from summary and confint.
verbose	A logical that indicates whether descriptive labels should be printed from summary and if certain warnings are shown with confint.
level	Note used, but her for compatability with generic confint function.
	Additional arguments for methods.

Details

The main function computes the estimates and associated standard errors, if possible, for the initial population size, No, and probability of capture, p, for seven methods chosen with method=. The possible methods are:

- method="CarleStrub": The general weighted k-pass estimator proposed by Carle and Strub (1978). This function iteratively solves for No in equation 7 of Carle and Strub (1978).
- method="Zippin": The general k-pass estimator generally attributed to Zippin. This function iteratively solves for No in bias corrected version of equation 3 (page 622) of Carle and Strub (1978). These results are not yet trustworthy.
- method="Seber3": The special case for k=3 estimator shown in equation 7.24 of Seber(2002).

- method="Seber2": The special case for k=2 estimator shown on page 312 of Seber(2002).
- method="RobsonRegier2": The special case for k=2 estimator shown by Robson and Regier (1968).
- method="Moran": The likelihood method of Moran (1951) as implemented by Schnute (1983).
- method="Schnute": The likelihood method of Schnute (1983) for the model that has a different probability of capture for the first sample but a constant probability of capture for all ensuing samples..

Confidence intervals for the first five methods are computed using standard large-sample normal distribution theory. Note that the confidence intervals for the 2- and 3-pass special cases are only approximately correct if the estimated population size is greater than 200. If the estimated population size is between 50 and 200 then a 95% CI behaves more like a 90% CI.

Confidence intervals for the last two methods use likelihood ratio theory as described in Schnute (1983) and are only produced for the No parameter. Standard errors are not produced with the Moran or Schnute methods..

In the Carle Strub method, if the resultant No estimate is equal to the sum of the catches (T) then the estimate of No that is returned will be the sum of the catches. In this instance, and if the "Seber" method of computing the standard error is used, then the SE will not be estimable and the confidence intervals can not be constructed.

Value

A vector that contains the estimates and standard errors for No and p if just.ests=TRUE or (default) a list with at least the following items:

- catch The original vector of observed catches.
- method The method used (provided by the user).
- lbl A descriptive label for the method used.
- est A matrix that contains the estimates and standard errors for No and p.

In addition, if the Moran or Schnute methods are used the list will also contain

- min.nlogLH The minimum value of the negative log-likelihood function.
- Tmult The Tmult value sent by the user.

testing

The Carle-Strub method matches the examples in Carle and Strub (1978) for No, p, and the variance of No. The Carle-Strub estimates of No and p match the examples in Cowx (1983) but the SE of No does not. The Carle-Strub estimates of No match the results (for estimates that they did not reject) from Jones and Stockwell (1995) to within 1 individual in most instances and within 1% for all other instances (e.g., off by 3 individuals when the esitmate was 930 individuals).

The Seber3 results for No match the results in Cowx (1983).

The Seber2 results for No, p, and the SE of No match the results in example 7.4 of Seber (2002) and in Cowx (1983).

The RobsonRegier2 results for No and the SE of NO match the resultsin Cowx (1983)

The Zippin method results do not match the examples in Seber (2002) or Cowx (1983) because removal uses the bias-corrected version from Carle and Strub (1978) and does not use the tables in Zippin (1958). The Zippin method is not yet trustworthy.

The Moran and Schnute methods match the examples in Schnute (1983) perfectly for all point estimates and within 0.1 units for all confidence intervals.

IFAR Chapter

10-Abundance from Depletion Data.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Carle, F.L. and M.R. Strub. 1978. A new method for estimating population size from removal data. Biometrics, 34:621-630.

Cowx, I.G. 1983. Review of the methods for estimating fish population size from survey removal data. Fisheries Management, 14:67-82.

Moran, P.A.P. 1951. A mathematical theory of animal trapping. Biometrika 38:307-311.

Robson, D.S., and H.A. Regier. 1968. Estimation of population number and mortality rates. pp. 124-158 in Ricker, W.E. (editor) Methods for Assessment of Fish Production in Fresh Waters. IBP Handbook NO. 3 Blackwell Scientific Publications, Oxford.

Schnute, J. 1983. A new approach to estimating populations by the removal method. Canadian Journal of Fisheries and Aquatic Sciences, 40:2153-2169.

Seber, G.A.F. 2002. The Estimation of Animal Abundance. Edward Arnold, second edition (Reprint).

van Dishoeck, P. 2009. Effects of catchability variation on performance of depletion estimators: Application to an adaptive management experiment. Masters Thesis, Simon Fraser University.

See Also

See depletion for related functionality.

```
## First example -- 3 passes
ct3 <- c(77,50,37)

# Carle Strub (default) method
p1 <- removal(ct3)
summary(p1)
summary(p1,verbose=TRUE)
summary(p1,parm="No")
summary(p1,parm="p")
confint(p1)
confint(p1,parm="No")</pre>
```

```
confint(p1,parm="p")
# Moran method
p2 <- removal(ct3,method="Moran")</pre>
summary(p2,verbose=TRUE)
confint(p2,verbose=TRUE)
# Schnute method
p3 <- removal(ct3,method="Schnute")
summary(p3,verbose=TRUE)
confint(p3,verbose=TRUE)
## Second example -- 2 passes
ct2 <- c(77,37)
# Seber method
p4 <- removal(ct2,method="Seber2")
summary(p4,verbose=TRUE)
confint(p4)
### Test if catchability differs between first sample and the other samples
# chi-square test statistic from negative log-likelihoods
# from Moran and Schnute fits (from above)
chi2.val <- 2*(p2$min.nlogLH-p3$min.nlogLH)</pre>
# p-value ... no significant difference
pchisq(chi2.val,df=1,lower.tail=FALSE)
# Another LRT example ... sample 1 from Schnute (1983)
ct4 <- c(45,11,18,8)
p2a <- removal(ct4,method="Moran")</pre>
p3a <- removal(ct4,method="Schnute")
chi2.val <- 2*(p2a$min.nlogLH-p3a$min.nlogLH) # 4.74 in Schnute(1983)</pre>
pchisq(chi2.val,df=1,lower.tail=FALSE) # significant difference (catchability differs)
summary(p3a)
### Using lapply() to use removal() on many different groups
### with the removals in a single variable ("long format")
## create a dummy data frame
lake <- factor(rep(c("Ash Tree", "Bark", "Clay"), each=5))</pre>
year <- factor(rep(c("2010","2011","2010","2011","2010","2011"),times=c(2,3,3,2,2,3)))</pre>
pass <- factor(c(1,2,1,2,3,1,2,3,1,2,1,2,1,2,3))
catch < c(57,34,65,34,12,54,26,9,54,27,67,34,68,35,12)
d <- data.frame(lake,year,pass,catch)</pre>
## create a variable that indicates each different group
d$group <- with(d,interaction(lake,year))</pre>
## split the catch by the different groups (creates a list of catch vectors)
ds <- split(d$catch,d$group)</pre>
## apply removal() to each catch vector (i.e., different group)
```

```
res <- lapply(ds,removal,just.ests=TRUE)</pre>
res <- data.frame(t(data.frame(res,check.names=FALSE)))</pre>
## get rownames from above and split into separate columns
nms <- t(data.frame(strsplit(rownames(res),"\\.")))</pre>
attr(nms, "dimnames") <- NULL</pre>
fnl <- data.frame(nms,res)</pre>
## put names together with values
rownames(fnl) <- NULL</pre>
colnames(fnl)[1:2] <- c("Lake", "Year")</pre>
fn1
### Using apply() to use removal() on many different groups
### with the removals in several variables ("wide format")
## create a dummy data frame (just reshaped from above as
## an example; -5 to ignore the group variable from above)
d1 <- reshape(d[,-5],timevar="pass",idvar=c("lake","year"),direction="wide")</pre>
## apply restore() to each row of only the catch data
res1 <- apply(d1[,3:5],MARGIN=1,FUN=removal,method="CarleStrub",just.ests=TRUE)
res1 <- data.frame(t(data.frame(res1,check.names=FALSE)))</pre>
## add the grouping information to the results
fnl1 <- data.frame(d1[,1:2],res1)</pre>
## put names together with values
rownames(fnl1) <- NULL</pre>
fnl1
```

residPlot

Construct a residual plot from lm or nls objects.

Description

Constructs a residual plot for 1m or nls objects. Different symbols for different groups can be added to the plot if an indicator variable regression is used.

Usage

```
residPlot(object, ...)
## S3 method for class 'lm'
residPlot(object, ...)
## S3 method for class 'SLR'
residPlot(object, xlab = "Fitted Values", ylab = "Residuals",
    main = NULL, pch = 16, col = "black", lty.ref = 3, lwd.ref = 1,
    col.ref = "black", resid.type = c("raw", "standardized", "studentized"),
    outlier.test = TRUE, alpha = 0.05, loess = TRUE, lty.loess = 2,
    lwd.loess = 1, col.loess = "black", trans.loess = 4, inclHist = TRUE,
    ...)
```

```
## S3 method for class 'POLY'
residPlot(object, ...)
## S3 method for class 'IVR'
residPlot(object, xlab = "Fitted Values", ylab = "Residuals",
 main = NULL, pch = c(16, 21, 15, 22, 17, 24, c(3:14)), col = "rich",
 lty.ref = 3, lwd.ref = 1, col.ref = "black", resid.type = c("raw",
 "standardized", "studentized"), outlier.test = TRUE, alpha = 0.05,
 loess = TRUE, lty.loess = 2, lwd.loess = 1, col.loess = "black",
 trans.loess = 4, legend = "topright", inclHist = TRUE, ...)
## S3 method for class 'ONEWAY'
residPlot(object, xlab = "Fitted Values",
 ylab = "Residuals", main = NULL, pch = 16, col = "black",
 lty.ref = 3, lwd.ref = 1, col.ref = "black", resid.type = c("raw",
 "standardized", "studentized"), bp = TRUE, outlier.test = TRUE,
 alpha = 0.05, loess = TRUE, lty.loess = 2, lwd.loess = 1,
 col.loess = "black", trans.loess = 4, inclHist = TRUE, ...)
## S3 method for class 'TWOWAY'
residPlot(object, xlab = "Fitted Values",
 ylab = "Residuals", main = NULL, pch = 16, col = "black",
 lty.ref = 3, lwd.ref = 1, col.ref = "black", resid.type = c("raw",
 "standardized", "studentized"), bp = TRUE, outlier.test = TRUE,
 alpha = 0.05, loess = TRUE, lty.loess = 2, lwd.loess = 1,
 col.loess = "black", trans.loess = 4, inclHist = TRUE, ...)
## S3 method for class 'nls'
residPlot(object, xlab = "Fitted Values", ylab = "Residuals",
 main = "", pch = 16, col = "black", lty.ref = 3, lwd.ref = 1,
 col.ref = "black", resid.type = c("raw", "standardized", "studentized"),
 loess = TRUE, lty.loess = 2, lwd.loess = 1, col.loess = "black",
 trans.loess = 4, inclHist = TRUE, ...)
## S3 method for class 'nlme'
residPlot(object, xlab = "Fitted Values", ylab = "Residuals",
 main = "", pch = 16, col = "black", lty.ref = 3, lwd.ref = 1,
 col.ref = "black", resid.type = c("raw", "standardized", "studentized"),
 loess = TRUE, lty.loess = 2, lwd.loess = 1, col.loess = "black",
 trans.loess = 4, inclHist = TRUE, ...)
```

Arguments

object	An lm or nls object (i.e., returned from fitting a model with either lm or nls).
xlab	A string for labelling the x-axis.
ylab	A string for labelling the y-axis.
main	A string for the main label to the plot. Defaults to the model call.

pch	A numeric that indicates the plotting charachter to be used or a vector of numerics that indicates what plotting charachters codes to use for the levels of the second factor. See par.
col	A vector of color names that indicates what color of points and lines to use for the levels of the first factor. See par.
lty.ref	A numeric that indicates the line type to use for the reference line at residual=0. See par.
lwd.ref	A numeric that indicates the line width to use for the reference line at residual=0. See par.
col.ref	A numeric or character that indicates the line color to use for the reference line at residual=0. See par.
resid.type	The type of residual to use. 'Raw' residuals are used by default. See details.
outlier.test	A logical that indicates if an outlierTest will TRUE (default) be performed and if the indivdiual with the largest studentized residual is deemed to be a significant outlier it will be noted on the residual plot by its observation number.
alpha	A numeric that indicates the alpha level to use for the outlier test (only used if outlier.test=TRUE).
loess	A logical that indicates if a loess smoother line and approximate confidence interval band is fit to and shown on the residual plot (TRUE).
lty.loess	A numeric that indicates the line type to use for loess fit line. See par.
lwd.loess	A numeric that indicates the line width to use for loess fit line. See par.
col.loess	A numeric or character that indicates the line color to use for loess fit line. See par.
trans.loess	A single numeric that indicates how transparent the loess band should be (larger numbers are more transparent).
inclHist	A logical that indicates if a second pane that includes the histogram of residuals should be constructed.
legend	If TRUE, draw a legend and the user must click in the upper-left corner of where the legend should be placed; if FALSE do not draw a legend. If a vector of length 2 then draw the upper left corner of the legend at the coordinates given in the vector of length 2.
bp	A logical that indicates if the plot for the one-way and two-way ANOVA will be a boxplot (TRUE; default) or not.
	Other arguments to the generic plot function.

Details

If the user chooses to add a legend without identifying coordinates for the upper-left corner of the legend (i.e., legend=TRUE) then the R console is suspended until the user places the legend by clicking on the produced graphic at the point where the upper-left corner of the legend should appear. A legend will only be placed if the mdl is an indicator variable regression, even if legend=TRUE.

Three types of residuals are allowed for most model types. Raw residuals are simply the difference between the observed response variable and the predicted/fitted value. Standardized residuals are

internally studentized residuals returned by rstandard for linear models and are the raw residual divided by the standard deviation of the residuals for nonlinear models (as is done by nlsResiduals from nlstools). Studentized residuals are the externally studentized residuals returned by rstudent for linear models and are not available for nonlinear models.

Externally Studentized residuals are not supported for nls or nlme objects. The

If outlier.test=TRUE then significant outliers are detected with outlierTest from the **car** package. See the help for this function for more details.

Value

None. However, a residual plot is produced.

Note

This function is meant to allow newbie students the ability to easily construct residual plots for one-way ANOVA, two-way ANOVA, simple linear regression, and indicator variable regressions. The plots can be constructed by submitting a saved linear model to this function which allows students to interact with and visualize moderately complex linear models in a fairly easy and efficient manner.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

See Also

See residualPlots in **car** and nlsResiduals in **nlstools**) for similar functionality and fitPlot and outlierTest in **car** for related methods.

```
data(Mirex)
Mirex$year <- factor(Mirex$year)</pre>
## Indicator variable regression with two factors
lm1 <- lm(mirex~weight*year*species,data=Mirex)</pre>
# defaults
residPlot(lm1)
# remove the histogram
residPlot(lm1,inclHist=FALSE)
# remove the loess line
residPlot(lm1,loess=FALSE,inclHist=FALSE)
# modify colors used
residPlot(lm1,col="rainbow",inclHist=FALSE)
# use only one point type -- notice that all points are of same type
residPlot(lm1,pch=16,inclHist=FALSE)
# use only one point and one color (might as well not use legend also)
residPlot(lm1,pch=16,col="black",legend=FALSE,inclHist=FALSE)
# can accomplish same thing just by removing the legend
residPlot(lm1,legend=FALSE,inclHist=FALSE)
# modify the reference line
residPlot(lm1,col.ref="blue",lwd.ref=5,inclHist=FALSE)
```

```
# use Studentized residuals
residPlot(lm1,resid.type="studentized",inclHist=FALSE)
# use Standardized residuals
residPlot(lm1,resid.type="standardized",inclHist=FALSE)
## Indicator variable regression with same two factors but in different order
     (notice use of colors and symbols)
lm1a <- lm(mirex~weight*species*year,data=Mirex)</pre>
residPlot(lm1a)
## Indicator variable regression with only one factor
lm2 <- lm(mirex~weight*year,data=Mirex)</pre>
residPlot(lm2)
## Indicator variable regression (assuming same slope)
lm3 <- lm(mirex~weight+year,data=Mirex)</pre>
residPlot(lm3)
## Simple linear regression
lm4 <- lm(mirex~weight,data=Mirex)</pre>
residPlot(lm4)
## One-way ANOVA
lm5 <- lm(mirex~year,data=Mirex)</pre>
# default (uses boxplots)
residPlot(lm5)
# use points rather than boxplot
residPlot(lm5,bp=FALSE)
## Two-Way ANOVA
lm6 <- lm(mirex~species*year,data=Mirex)</pre>
# default (uses boxplots)
residPlot(lm6)
# No boxplots
residPlot(lm6,bp=FALSE)
## Example showing outlier detection
x \leftarrow c(runif(100))
y <- c(7,runif(99))</pre>
lma <- lm(y~x)
residPlot(lma)
# with studentized residuals
residPlot(lm1,resid.type="studentized")
## Nonlinear regression
# from first example in nls()
DNase1 <- subset(DNase,Run==1)</pre>
fm1DNase1 <- nls(density~SSlogis(log(conc),Asym,xmid,scal),DNase1)</pre>
residPlot(fm1DNase1)
residPlot(fm1DNase1,resid.type="standardized")
```

Schnute 161

Schnute

The four-parameter growth function from Schnute (1981).

Description

The four-parameter growth function from Schnute (1981). Use SchnuteModels() to see the equations for each growth function.

Usage

```
Schnute(t, case = 1, t1 = NULL, t3 = NULL, L1 = NULL, L3 = NULL,
    a = NULL, b = NULL)
SchnuteModels(cex = 1.25, ...)
```

Arguments

t	A numeric vector of ages over which to model growth.
case	A string that indicates the case of the Schnute growth function to use.
t1	The (young) age that corresponds to L1. Set to minimum value in t by default.
t3	The (old) age that corresponds to L3. Set to maximum value in t by default.
L1	The mean size/length at t1.
L3	The mean size/length at t3.
а	A dimensionless parameter that is related to the time/age at the inflection point.
b	A dimensionless parameter that is related to size/length at the inflection point.
cex	A single numeric expansion value for use with SchnuteModels.
	Not implemented.

Value

Schnute returns a predicted size given the case of the function and the provided parameter values. SchnuteModels returns a graphic that uses plotmath to show the growth function equation in a pretty format.

IFAR Chapter

None specifically, but 12-Individual Growth is related.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Schnute, J. 1981. A versatile growth model with statistical stable parameters. Canadian Journal of Fisheris and Aquatic Sciences 38:1128-1140.

162 se

See Also

See vbFuns, GompertzFuns, RichardsFuns, and logisticFuns for similar functionality for other models.

Examples

```
## See the formulae
## Not run: windows(5,5)
SchnuteModels()

## Simple examples
ages <- 1:15
s1 <- Schnute(ages,case=1,t1=1,t3=15,L1=30,L3=400,a=0.3,b=1)
s2 <- Schnute(ages,case=2,t1=1,t3=15,L1=30,L3=400,a=0.3,b=1)
s3 <- Schnute(ages,case=3,t1=1,t3=15,L1=30,L3=400,a=0.3,b=1)
s4 <- Schnute(ages,case=4,t1=1,t3=15,L1=30,L3=400,a=0.3,b=1)

plot(s1~ages,type="1",lwd=2)
lines(s2~ages,lwd=2,col="red")
lines(s3~ages,lwd=2,col="blue")
lines(s4~ages,lwd=2,col="green")</pre>
```

se

Computes standard error of the mean.

Description

Computes the standard error of the mean (i.e., standard deviation divided by the square root of the sample size).

Usage

```
se(x, na.rm = TRUE)
```

Arguments

x A numeric vector.

na.rm A logical that indicates whether missing values should be removed before computing the standard error.

Value

A single numeric that is the standard error of the mean of x.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

SMBassLS 163

See Also

See se in **sciplot** for similar functionality.

Examples

```
x <- 1:20
sd(x)/sqrt(length(x))
se(x)

# all return NA if missing values are not removed
x2 <- c(x,NA)
sd(x2)/sqrt(length(x2))

# Better if missing values are removed
se(x2,na.rm=FALSE)
sd(x2,na.rm=TRUE)/sqrt(length(x2[complete.cases(x2)]))
se(x2)</pre>
```

SMBassLS

Catch-effort data for Little Silver Lake Smallmouth Bass.

Description

Catch-effort data for Smallmouth Bass (Micropterus dolomieu) in Little Silver Lake, ONT.

Format

A data frame with 10 observations on the following 3 variables:

day Day of the catch.

catch Number of smallmouth bass caught.

effort Number of traps set per day.

Topic(s)

- Population size
- Abundance
- · Depletion methods
- · Leslie method
- · DeLury method
- · Catchability

Source

From Omand, D.N. 1951. A study of populations of fish based on catch-effort statistics. Journal of Wildlife Management, 15:88-98.

164 SMBassWB

Examples

data(SMBassLS)
str(SMBassLS)
head(SMBassLS)

SMBassWB

Growth increment data for West Bearskin Lake Smallmouth Bass.

Description

Growth data from Smallmouth Bass (*Micropterus dolomieu*) captured in West Bearksin Lake, MN. Five samples were collected over three years (1988-1990) with two gears (fall – trapnets, spring – electrofishing).

Format

A data frame of 445 observations on the following 20 variables:

species Species of the fish (SMB for each fish in this file).

lake Lake fish was captured in (WB for each fish in this file).

gear Gear used to capture the fish (T=Trapnet and E=Electrofishing).

yearcap Year fish was captured (1988, 1989, or 1990).

fish A unique identifier for each fish.

agecap Assigned age-at-capture for the fish (from scales).

lencap Total length-at-capture for the fish (mm).

anu1 Magnified scale radius (mm) to the 1st annulus.

anu2 Magnified scale radius (mm) to the 2md annulus.

anu3 Magnified scale radius (mm) to the 3rd annulus.

anu4 Magnified scale radius (mm) to the 4th annulus.

anu5 Magnified scale radius (mm) to the 5th annulus.

anu6 Magnified scale radius (mm) to the 6th annulus.

anu7 Magnified scale radius (mm) to the 7th annulus.

anu8 Magnified scale radius (mm) to the 8th annulus.

anu9 Magnified scale radius (mm) to the 9th annulus.

anu10 Magnified scale radius (mm) to the 10th annulus.

anu11 Magnified scale radius (mm) to the 11th annulus.

anu12 Magnified scale radius (mm) to the 12th annulus.

radcap Total scale radius at time of capture.

SpotVA1

Topic(s)

- Growth increment analysis
- · Weisberg linear growth model

Note

Data are in one-fish-per-line format.

Source

Data from the linear growth modelling software distributed in support of Weisberg, S. 1993. Using hard-part increment data to estimate age and environmental effects. Canadian Journal of Fisheries and Aquatic Sciences 50:1229-1237.

See Also

See wblake2 from alr3 for the same dataset with only the agecap, lencap, and radcap variables.

Examples

```
data(SMBassWB)
str(SMBassWB)
head(SMBassWB)
```

SpotVA1

Age and length of spot.

Description

Ages (from otoliths) and length of Virginia Spot (Leiostomus xanthurus).

Format

A data frame of 403 observations on the following 2 variables:

tl Measured total lengths (in inches).

age Ages assigned from examination of otoliths.

Details

Final length measurements were simulated by adding a uniform error to the value at the beginning of the length category.

Topic(s)

- Growth
- · von Bertalanffy

166 srStarts

Source

Extracted from Table 1 in Chapter 8 (Spot) of the VMRC Final Report on Finfish Ageing, 2002 by the Center for Quantitative Fisheries Ecology at Old Dominion University.

See Also

See SpotVA2 in **FSAdata** for related data.

Examples

```
data(SpotVA1)
str(SpotVA1)
head(SpotVA1)
plot(tl~age,data=SpotVA1)
```

srStarts

Finds reasonable starting values for parameters in specific paramaterizations of common stock-recruitment models.

Description

Finds reasonable starting values for parameters in specific parameterizations of the "Beverton-Holt", "Ricker", "Shepherd", or "Saila-Lorda" stock-recruitment models. Type srModels() for the equations of each model.

Usage

```
srStarts(formula, data = NULL, type = c("BevertonHolt", "Ricker",
   "Shepherd", "SailaLorda", "independence"), param = 1, plot = FALSE,
   col.mdl = "gray70", lwd.mdl = 3, lty.mdl = 1, dynamicPlot = FALSE,
   min.prop = 0.1, max.mult = 3, delta.prop = 0.005, ...)
```

Arguments

formula	A formula of the form Recruits~Stock.
data	A data frame in which Recruits and Stock are found.
type	A string that indicates the type of the stock-recruitment model. Must be one of "BevertonHolt", "Ricker", "Shepherd", or "SailaLorda".
param	A numeric that indicates the parameterization of the stock-recruitment model type. This is ignored if $type="Shepherd"$ or $type="SailaLorda"$
plot	A logical that indicates whether or not a plot of the data with the model fit at the starting values superimposed is created.
col.mdl	A color for the model when plot=TRUE.
lwd.mdl	A line width for the model when plot=TRUE.
lty.mdl	A line type for the model when plot=TRUE.

srStarts 167

dynamicPlot A logical that indicates where a plot with dynamically linked slider bars should be constructed for finding starting values.

min.prop A single numeric that is used to set the minimum values for the slider bars in the dynamic plot. see details.

A single numeric that is used to set the maximum values for the slider bars in the dynamic plot. see details.

A single numeric that is used to set the step value for the slider bars in the dynamic plots. See details.

Further arguments passed to the methods.

Details

This function attempts to find reasonable starting values for a variety of parameterizations of the "Beverton-Holt", "Ricker", "Shepherd", or "Saila-Lorda" stock-recruitment models. There is no guarantee that these starting values are the 'best' starting values. One should use them with caution and should perform sensitivity analyses to determine the impact of different starting values on the final model results.

Starting values for the first parameterization of the Beverton-Holt model were derived by linearizing the function (inverting both sides and simplifying), fitting a linear model to the observed data, and extracting parameter values from the corresponding linear model parameters. Starting values for the other parameterizations of the Beverton-Holt model were derived from known relationships between the parameters of each parameterization and the first parameterization. If the computed starting value for the Rp parameter was larger than the largest observed recruitment value, then the starting value for Rp was set to the largest observed recruitment value.

Starting values for the Shepherd function were the same as those for the first parameterization of the Beverton-Holt function with the addition that c=1.

Starting values for the Ricker parameterizations followed the same general procedure as described for the Beverton-Holt parameterizations. If the computed starting value for atilde was less than zero then the starting value was set to 0.00001.

Starting values for the Saila-Lorda function were the same as those for the first parameterization of the Ricker function with the addition that c=1.

Starting values can also be obtained by plotting the data with the model superimposed but tied to slider bars for changing parameters. One can change the parameters until a reasonable fit is observed and then use those valeus as starting values. Such a plot can be constructed by including dynamicPlot=TRUE. The initial parameters for the slider bars are the starting values constructed as described above. The range for the sliders will have a minimum that is min, prop times the initial value and a maximum that is max.mult times the initial value. The step or interval of the slider bar is delta.mult times the initial value. It should be noted that the dynamic plot may show an error of "[tcl] can't get device image" but the plot will correctly update if the slider bar is adjusted.

Value

A list that contains reasonable starting values. Note that the parameters will be listed in the same order and with the same names as listed in srFuns.

168 srStarts

IFAR Chapter

13-Recruitment.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Beverton, R.J.H. and S.J. Holt. 1957. On the dynamics of exploited fish populations, Fisheries Investigations (Series 2), volume 19. United Kingdom Ministry of Agriculture and Fisheries, 533 pp.

Iles, T.C. 1994. A review of stock-recruitment relationships with reference to flatfish populations. Netherlands Journal of Sea Research 32:399-420.

Quinn II, T.J. and R.B. Deriso. 1999. Quantitative Fish Dynamics. Oxford University Press.

Ricker, W.E. 1954. Stock and recruitment. Journal of the Fisheries Research Board of Canada 11:559-623.

Ricker, W.E. 1975. Computation and interpretation of biological statistics of fish populations. Technical Report Bulletin 191, Bulletin of the Fisheries Research Board of Canada.

Shepherd, J. 1982. A versatile new stock-recruitment relationship for fisheries and construction of sustainable yield curves. Journal du Conseil International pour l'Exploration de la Mar 40:67-75.

See Also

See srModels and srFuns for related functionality.

```
## Simple Examples
data(CodNorwegian)
srStarts(recruits~stock,data=CodNorwegian)
srStarts(recruits~stock,data=CodNorwegian,param=2)
srStarts(recruits~stock,data=CodNorwegian,param=3)
srStarts(recruits~stock,data=CodNorwegian,param=4)
srStarts(recruits~stock,data=CodNorwegian,type="Ricker")
srStarts(recruits~stock,data=CodNorwegian,type="Ricker",param=2)
srStarts(recruits~stock,data=CodNorwegian,type="Ricker",param=3)
srStarts(recruits~stock,data=CodNorwegian,type="Shepherd")
srStarts(recruits~stock,data=CodNorwegian,type="SailaLorda")
srStarts(recruits~stock,data=CodNorwegian,type="independence")
## Simple Examples with a Plot
srStarts(recruits~stock,data=CodNorwegian,type="Ricker",plot=TRUE)
srStarts(recruits~stock,data=CodNorwegian,type="BevertonHolt",plot=TRUE)
srStarts(recruits~stock,data=CodNorwegian,type="Shepherd",plot=TRUE)
srStarts(recruits~stock,data=CodNorwegian,type="SailaLorda",plot=TRUE)
srStarts(recruits~stock,data=CodNorwegian,type="independence",plot=TRUE)
```

stockRecruitment 169

```
## Dynamic Plots Method -- ONLY RUN IN INTERACTIVE MODE
if (interactive()) {
 # Beverton-Holt
 srStarts(recruits~stock,data=CodNorwegian,dynamicPlot=TRUE)
 srStarts(recruits~stock,data=CodNorwegian,param=2,dynamicPlot=TRUE)
 srStarts(recruits~stock,data=CodNorwegian,param=3,dynamicPlot=TRUE)
 srStarts(recruits~stock,data=CodNorwegian,param=4,dynamicPlot=TRUE)
 # Ricker Models
 srStarts(recruits~stock,data=CodNorwegian,type="Ricker",dynamicPlot=TRUE)
 srStarts(recruits ``stock, data = CodNorwegian, type = "Ricker", param = 2, dynamicPlot = TRUE)
 srStarts(recruits ``stock, data=CodNorwegian, type="Ricker", param=3, dynamicPlot=TRUE)
 # Shepherd, Saila-Lorda, and Independence Models
 srStarts(recruits~stock,data=CodNorwegian,type="Shepherd",dynamicPlot=TRUE)
 srStarts(recruits~stock,data=CodNorwegian,type="SailaLorda",dynamicPlot=TRUE)
 srStarts(recruits ``stock, data = CodNorwegian, type = "independence", dynamicPlot = TRUE)
} ## END .. ONLY INTERACTIVE
## See examples in srFuns() for use of srStarts() when fitting stock-recruit models
```

stockRecruitment

Creates a function for a specific parameterization of a common stock-recruitment model .

Description

Creates a function for a specific parameterization of a "Beverton-Holt", "Ricker", "Shepherd", or "Saila-Lorda" stock-recruitment model. Use srModels() to see the equations of each model.

Usage

```
srFuns(type = c("BevertonHolt", "Ricker", "Shepherd", "SailaLorda",
    "independence"), param = 1, simple = FALSE, msg = FALSE)
srModels(...)
```

Arguments

type	A string that indicates the type of stock-recruitment model.
param	A single numeric that indicates the parameterization of the stock-recruitment model type.
simple	A logical that indicates whether the user should be allowed to send all parameter values in the first parameter argument (=FALSE; default) or whether all individual parameters must be specified (=TRUE).
msg	A logical that indicates whether a message about the model and parameter definitions should be output (=TRUE) or not (=FALSE; default).
	Not implemented.

170 stockRecruitment

Value

srFuns returns a function that can be used to predict recruitment given a vector of stock sizes and values for the model parameters. The result should be saved to an object that can then be used as a function name. When the resulting function is used, the parameters are ordered as shown when the definitions of the parameters are printed after the function is called (assuming that msg=TRUE). The values for both/all parameters can be included as a vector of length two/three in the first parameter argument. If simple=FALSE then the values for all parameters can be included as a vector in the first parameter argument. If simple=TRUE then all parameters must be declared individually in each function. The resulting function is somewhat easier to read when simple=TRUE.

srModels returns a graphic that uses plotmath to show the model formulae in a pretty format.

IFAR Chapter

13-Recruitment.

Author(s)

Derek H. Ogle, <derek@derekogle.com>, thanks to Gabor Grothendieck for a hint about using get().

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

Beverton, R.J.H. and S.J. Holt. 1957. On the dynamics of exploited fish populations, Fisheries Investigations (Series 2), volume 19. United Kingdom Ministry of Agriculture and Fisheries, 533 pp.

Iles, T.C. 1994. A review of stock-recruitment relationships with reference to flatfish populations. Netherlands Journal of Sea Research 32:399-420.

Quinn II, T.J. and R.B. Deriso. 1999. Quantitative Fish Dynamics. Oxford University Press.

Ricker, W.E. 1954. Stock and recruitment. Journal of the Fisheries Research Board of Canada 11:559-623.

Ricker, W.E. 1975. Computation and interpretation of biological statistics of fish populations. Technical Report Bulletin 191, Bulletin of the Fisheries Research Board of Canada.

Shepherd, J. 1982. A versatile new stock-recruitment relationship for fisheries and construction of sustainable yield curves. Journal du Conseil International pour l'Exploration de la Mar 40:67-75.

See Also

See srStarts for related functionality.

```
## See the formulae
## Not run: windows(6,5)
srModels()
## Simple Examples
```

Subset 171

```
# show what a message looks like with the function definition
srFuns("Ricker",msg=TRUE)
# create some dummy stock data
stock <- seq(0.01,1000,length.out=199)
# Beverton-Holt #1 parameterization
( bh1 <- srFuns() )</pre>
plot(bh1(stock, a=0.5, b=0.01) \sim stock, type="l", lwd=2, ylab="Recruits", xlab="Spawners", ylim=c(0,50))
# Ricker #1 parameterization
( r1 <- srFuns("Ricker") )</pre>
lines(r1(stock, a=0.5, b=0.005)~stock, lwd=2, col="red")
# Shephered parameterization
( s1 <- srFuns("Shepherd") )</pre>
lines(s1(stock,a=0.5,b=0.005,c=2.5)~stock,lwd=2,col="blue")
# Saila-Lorda parameterization
( sl1 <- srFuns("SailaLorda") )</pre>
lines(sl1(stock,a=0.5,b=0.005,c=1.05)~stock,lwd=2,col="salmon")
## Examples of fitting stock-recruitment models
data(CodNorwegian)
# Fitting the Beverton-Holt #1 parameterization with multiplicative errors
bh1s <- srStarts(recruits~stock,data=CodNorwegian)</pre>
fit1 <- nls(log(recruits)~log(bh1(stock,a,b)),data=CodNorwegian,start=bh1s)</pre>
summary(fit1,correlation=TRUE)
\verb|plot(recruits~stock,data=CodNorwegian,pch=19,xlim=c(0,200))||
curve(bh1(x,a=coef(fit1)[1],b=coef(fit1)[2]),from=0,to=200,col="red",lwd=3,add=TRUE)\\
# Fitting the Ricker #3 parameterization with multiplicative errors
r3 <- srFuns("Ricker",param=3)
r3s <- srStarts(recruits~stock,data=CodNorwegian,type="Ricker",param=3)
fit2 <- nls(log(recruits) \sim log(r3(stock,a,Rp)), data = CodNorwegian, start = r3s)
summary(fit2,correlation=TRUE)
curve(r3(x,a=coef(fit2)[1],Rp=coef(fit2)[2]),from=0,to=200,col="blue",lwd=3,add=TRUE)
```

Subset

Subsets/filters a data frame and drops the unused levels.

Description

Subsets/filters a data frame and drops the unused levels.

Usage

```
Subset(x, subset, select, drop = FALSE, resetRownames = TRUE, ...)
filterD(x, ...)
```

Subset Subset

Arguments

x A data frame.

subset A logical expression that indicates elements or rows to keep: missing values are

taken as false.

select An expression, that indicates columns to select from a data frame.

drop passed on to [indexing operator.

resetRownames A logical that indicates if the rownames should be reset after the subsetting

(TRUE; default). Resetting rownames will simply number the rows from 1 to the

number of rows in the result.

. . . further arguments to be passed to or from other methods.

Details

Newbie students using R expect that when a factor variable is subsetted with subset or filtered with filter that any original levels that are no longer used after the subsetting or filtering will be ignored. This, however, is not the case and often results in tables with empty cells and figures with empty bars. One remedy is to use drop.levels from gdata immediately following the subset or filter call. This generally becomes a repetitive sequence for most newbie students; thus, Subset and filterD incorporate these two functions into one function.

Subset is a wrapper to subset with a catch for non-data.frames and a specific call to drop.levels just before the data.frame is returned. I also added an argument to allow resetting the row names. filterD is a wrapper for filter from **dplyr** followed by drop.levels just before the data.frame is returned. Otherwise, there is no new code here.

These functions are used only for data frames.

Value

A data frame with the subsetted rows and selected variables.

IFAR Chapter

Basic Data Manipulations.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

See Also

See subset and filter from **dplyr** for similar functionality. See drop.levels in **gdata** and droplevels for related functionality.

Summarize 173

Examples

```
## The problem -- note use of unused level in the final table.
levels(iris$Species)
iris.set1 <- subset(iris,Species=="setosa" | Species=="versicolor")
levels(iris.set1$Species)
xtabs(~Species,data=iris)

## A simpler fix using Subset
iris.set2 <- Subset(iris,Species=="setosa" | Species=="versicolor")
levels(iris.set2$Species)
xtabs(~Species,data=iris.set2)

## A simpler fix using filterD
iris.set3 <- filterD(iris,Species=="setosa" | Species=="versicolor")
levels(iris.set3$Species)
xtabs(~Species,data=iris.set3)</pre>
```

Summarize

Summary statistics for a numeric or factor variable.

Description

Summary statistics for a single numeric or factor variable, possibly separated by the levels of a factor variable. This function is very similar to summary for a numeric variables and table for factor variables.

Usage

```
Summarize(object, ...)
## Default S3 method:
Summarize(object, digits = getOption("digits"),
   addtotal = TRUE, percent = c("total", "none"), percdigs = 2,
   na.rm = TRUE, exclude = "", ...)
## S3 method for class 'formula'
Summarize(object, data = NULL,
   digits = getOption("digits"), percent = c("row", "column", "total",
   "none"), percdigs = 2, addtotal = TRUE, na.rm = TRUE, exclude = "",
   ...)
```

Arguments

object A vector of numeric or factor data.

digits A numeric that indicates the number of decimals to round the numeric summaries.

addtotal A logical that indicates whether totals should be added to tables (=TRUE, default)

or not. See details.

174 Summarize

percent A string that indicates the type of percents to compute for tables from factor

variables. See details.

percdigs A numeric that indicates the number of decimals to round the percentage sum-

maries.

na.rm A logical that indicates whether numeric missing values (NA) should be removed

(=TRUE, default) or not.

exclude A string that contains the level that should be excluded from a factor variable.

data An optional data frame that contains the variables in formula.

... Not implemented.

Details

This function is primarily used with formulas. Five general types of formulae may be used (where quant and factor generically represent quantitative/numeric and factor variables, respectively)

Formula Description of Summary

~quant Numerical summaries (see below) of quant.

~factor One-way frequency or percentage (see below) table of factor.

quant~factor Summaries of quant separated by levels in factor.

quant~factor1*factor2 Summaries of quant separated by the combined levels in factor1 and factor2.

factor1~factor2 Two-way frequency or percentage table with levels of factor2 as rows and factor1 as columns.

Numerical summaries include all results from summary (min, Q1, mean, median, Q3, and max) and the sample size, valid sample size (sample size minus number of NAs), and standard deviation (i.e., sd). NA values are removed from the calculations with na.rm=TRUE (the DEFAULT). The number of digits in the returned results are controlled with digits=.

Factor variables may be summarized as a frequency (if percent="none") or percentages table (the DEFAULT). For a single factor variable, the percentages table is returned if percent="total". For two factor variables, the percentage table may be returned as a row-, column-, or table-percent table with percent="row" (the DEFAULT), percent="column", and percent="total", respectively. The number of digits in the returned table are controlled with percdigs=. A marginal total, either for all margins if percent="none" or the appropriate margin otherwise, is added to the table if addtotal=TRUE. The results for a factor are NOT meant to replace table or xtabs. This functionality is provided to make this function more complete.

Value

A named vector or data frame (when a quantitative variable is separted by one or two factor variables) of summary statistics for numeric data and a matrix of frequencies and, possibly, percentages for factor variables.

Note

Students often need to examine basic statistics of a quantitative variable separated for different levels of a categorical variable. These results may be obtained with tapply, by, or aggregate (or with functions in other packages), but the use of these functions is not obvious to newbie students

Summarize 175

or return results in a format that is not obvious to newbie students. Thus, the formula method to Summarize allows newbie students to use a common notation (i.e., formula) to easily compute summary statistics for a quantitative variable separated by the levels of a factor.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

See Also

See summary, table, and xtabs for related one dimensional functionality. See tapply, summaryBy in **doBy**, describe in **psych**, describe in **prettyR**, and basicStats in **fBasics** for similar "by" functionality.

```
## Create a numeric vector (with missing values)
n <- 102
y \leftarrow c(0,0,NA,NA,NA,runif(n-5))
## Create a factor vector (with missing values)
g1 <- factor(sample(c("A","B","C","NA"),n,replace=TRUE))</pre>
## Create a factor vector with unknowns
g2 <- factor(sample(c("male","female","UNKNOWN"),n,replace=TRUE))</pre>
# Put into a data.frame (with some extra variables)
d <- data.frame(dy=y,dg1=g1,dg2=g2,</pre>
                dw=sample(1:3,n,replace=TRUE),
                dv=sample(1:3,n,replace=TRUE))
# typical output of summary() for a numeric variable
summary(y)
# this function
Summarize(y,digits=3)
Summarize(~dy,data=d,digits=3)
Summarize(dy~1,data=d,digits=3)
## Factor vector (excluding "NA"s in second call)
Summarize(~dg1,data=d)
Summarize(~dg1,data=d,exclude="NA")
## Factor vector with UNKNOWNs
Summarize(~dg2,data=d)
Summarize(~dg2,data=d,exclude="UNKNOWN")
## Numeric vector by levels of a factor variable
Summarize(dy~dg1,data=d,digits=3)
Summarize(dy~dg1,data=d,digits=3,exclude="NA")
Summarize(dy~dg2,data=d,digits=3)
Summarize(dy~dg2,data=d,digits=3,exclude="UNKNOWN")
## What happens if RHS of formula is not a factor
Summarize(dy~dw,data=d,digits=3)
```

176 sumTable

```
Summarize(y~dw*dv,data=d,digits=3)
## Summarize factor variable by a factor variable
Summarize(dg1~dg2,data=d)
Summarize(dg1~dg2,data=d,exclude="NA")
Summarize(dg1~dg2,data=d,exclude=c("NA","UNKNOWN"))
Summarize(dg1~dg2,data=d,percent="none")
Summarize(dg1~dg2,data=d,percent="column")
Summarize(dg1~dg2,data=d,percent="total")
## Summarizing all variables in a data frame
lapply(as.list(d),Summarize,digits=4)
```

sumTable

Creates a one- or two-way table of summary statistics.

Description

Creates a one- or two-way table of summary statistics for a quantitative variable.

Usage

```
sumTable(formula, ...)
## S3 method for class 'formula'
sumTable(formula, data = NULL, FUN = mean,
    digits = getOption("digits"), ...)
```

Arguments

formula	A formula with a quantitative variable on the left-hand-side and one or two factor variables on the right-hand-side. See details.
data	An optional data frame that contains the variables in formula.
FUN	A scalar function that identifies the summary statistics. Applied to the quantitative variable for all data subsets identified by the combination of the factor(s). Defaults to mean.
digits	A single numeric that indicates the number of digits to be used for the result.
	Other arguments to pass through to FUN.

Details

The formula must be of the form quantitative~factor or quantitative~factor* where quantitative is the quantitative variable to construct the summaries for and factor and factor2 are factor variables that contain the levels for which separate summaries should be constructed. If the variables on the right-hand-side are not factors, then they will be coerced to be factors and a warning will be issued.

This function is largely a wrapper to tapply(), but only works for one quantitative variable on the left-hand-side and one or two factor variales on the right-hand-side. Consider using tapply for situations with more factors on the right-hand-side.

tictactoe 177

Value

A one-way array of values if only one factor variable is supplied on the right-hand-side of formula. A two-way matrix of values if two factor variables are supplied on the right-hand-side of formula. These are the same classes of objects returned by tapply.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

See Also

See tapply for a more general implementation. See Summarize for a similar computation when only one factor variable is given.

Examples

```
## The same examples as in the old aggregate.table in gdata package
## but data in data.frame to illustrate formula notation
d <- data.frame(g1=sample(letters[1:5], 1000, replace=TRUE),</pre>
                g2=sample(LETTERS[1:3], 1000, replace=TRUE),
                dat=rnorm(1000))
sumTable(dat~g1*g2,data=d,FUN=length)
                                            # get sample size
sumTable(dat~g1*g2,data=d,FUN=validn)
                                            # get sample size (better way)
sumTable(dat~g1*g2,data=d,FUN=mean)
                                            # get mean
sumTable(dat~g1*g2,data=d,FUN=sd)
                                            # get sd
sumTable(dat~g1*g2,data=d,FUN=sd,digits=1) # show digits= argument
## Also demonstrate use in the 1-way example -- but see Summarize() in FSA package
sumTable(dat~g1,data=d,FUN=validn)
sumTable(dat~g1,data=d,FUN=mean)
## Example with a missing value (compare to above)
d$dat[1] <- NA
sumTable(dat~g1,data=d,FUN=validn) # note use of validn
sumTable(dat~g1,data=d,FUN=mean,na.rm=TRUE)
```

tictactoe

Construct a base tic-tac-toe plot for presenting predator-prey PSD values.

Description

Construct a base tic-tac-toe plot for presenting predator-prey PSD values. Predator-prey PSD values are added with plotCI from **plotrix**.

178 tictactoe

Usage

```
tictactoe(predobj = c(30, 70), preyobj = c(30, 70),
predlab = "Predator PSD", preylab = "Prey PSD", obj.col = "black",
obj.trans = 0.2, bnd.col = "black", bnd.lwd = 1, bnd.lty = 2)
```

Arguments

predobj	A vector of length 2 that contains the target objective range for the predator.
preyobj	A vector of length 2 that contains the target objective range for the prey.
predlab	A string representing a label for the x-axis.
preylab	A string representing a label for the y-axis.
obj.col	A string designating a color to which the target objective regions should be shaded.
obj.trans	A numeric (decimal) that indicates the level of transparency for marking the target objective regions.
bnd.col	A string that indicates a color for the boundaries of the target objective regions.
bnd.lwd	A numeric that indicates the line width for the boundaries of the target objective regions.
bnd.lty	A numeric that indicates the line type for the boundaries of the target objective regions.

Details

This function simply creates a base tic-tac-toe plot. Observed values, with confidence intervals, are added to this plot with plotCI from **plotrix**; see examples.

Value

None. However, a graphic is produced.

IFAR Chapter

6-Size Structure.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

See Also

See psdVal and psdCalc for related functionality.

validn 179

Examples

```
## Create hypothetical data for plotting one point .. similar to what might come from psdCalc()
prey <- c(45.4,30.2,56.8)
pred <- c(24.5,10.2,36.7)
names(prey) <- names(pred) <- c("Estimate","95% LCI","95% UCI")</pre>
prey
pred
tictactoe()
if (require(plotrix)) {
  plotCI(prey[1],pred[1],li=prey[2],ui=prey[3],err="x",pch=16,add=TRUE)
  plotCI(prey[1],pred[1],li=pred[2],ui=pred[3],err="y",pch=16,add=TRUE)
}
## Create hypothetical data for plotting three points .. similar to what might come from psdCalc()
prey <- rbind(c(45.4,30.2,56.8),
              c(68.2,56.7,79.4),
              c(17.1, 9.5,26.3))
pred <- rbind(c(24.5,10.2,36.7),
              c(14.2, 7.1, 21.3),
              c(16.3, 8.2,24.4))
colnames(prey) <- colnames(pred) <- c("Estimate", "95% LCI", "95% UCI")</pre>
prey
pred
tictactoe()
if (require(plotrix)) {
  plotCI(prey[,1],pred[,1],li=prey[,2],ui=prey[,3],err="x",pch=16,add=TRUE)
  plotCI(prey[,1],pred[,1],li=pred[,2],ui=pred[,3],err="y",pch=16,add=TRUE)
lines(prey[,1],pred[,1])
text(prey[,1],pred[,1],labels=c(2010,2011,2012),adj=c(-0.5,-0.5))
```

validn

Finds the number of valid (non-NA) values in a vector.

Description

Finds the number of valid (non-NA) values in a vector.

Usage

```
validn(object)
```

Arguments

object

A vector.

vbStarts

Value

A single numeric value that is the number of non-NA values in a vector.

IFAR Chapter

2-Basic Data Manipulations.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

See Also

See valid.n in **plotrix** and nobs in **gdata** for similar functionality. See is.na for finding the missing values.

Examples

```
junk1 <- c(1,7,2,4,3,10,NA)
junk2 <- c("Derek","Hugh","Ogle","Santa","Claus","Nick",NA,NA)
junk3 <- factor(junk2)
junk4 <- c(TRUE,TRUE,FALSE,FALSE,TRUE,NA,NA)
junk5 <- data.frame(junk1)
junk6 <- data.frame(junk3)

validn(junk1)
validn(junk2)
validn(junk3)
validn(junk4)
validn(junk5)
validn(junk5)</pre>
```

vbStarts

Find reasonable starting values for a von Bertalanffy growth function.

Description

Finds reasonable starting values for the parameters in a specific parameterization of the von Bertalanffy growth function.

Usage

```
vbStarts(formula, data = NULL, type = c("Typical", "typical",
   "BevertonHolt", "Original", "original", "vonBertalanffy", "GQ",
   "GallucciQuinn", "Mooij", "Weisberg", "Schnute", "Francis", "Somers",
   "Somers2"), ages2use = NULL, methEV = c("means", "poly"),
   meth0 = c("yngAge", "poly"), plot = FALSE, col.mdl = "gray70",
   lwd.mdl = 3, lty.mdl = 1, cex.main = 0.75, dynamicPlot = FALSE, ...)
```

vbStarts 181

Arguments

formula	A formula of the form len~age.
data	A data frame that contains the variables in formula.
type	A string that indicates the parameterization of the von Bertalanffy model.
ages2use	A numerical vector of the two ages to be used in the Schnute or Francis paramaterizations. See details.
methEV	A string that indicates how the lengths of the two ages in the Schnute paramaterization or the three ages in the Francis paramaterization should be derived. See details.
meth0	A string that indicates how the t0 and L0 paramaters should be derived. See details.
plot	A logical that indicates whether a plot of the data with the superimposed model fit at the starting values should be created.
col.mdl	A color for the model when plot=TRUE.
lwd.mdl	A line width for the model when plot=TRUE.
lty.mdl	A line type for the model when plot=TRUE.
cex.main	A character expansion value for the main title when plot=TRUE.
dynamicPlot	A logical that indicates whether a plot with dynamically linked slider bars should be constructed for finding starting values.
	Further arguments passed to the methods.

Details

This function attempts to find reasonable starting values for a variety of parameterizations of the von Bertalanffy growth function. There is no guarantee that these starting values are the 'best' starting values. One should use them with caution and should perform sensitivity analyses to determine the impact of different starting values on the final model results.

The Linf and K paramaters are estimated via the concept of the Ford-Walford plot. The product of the starting values for Linf and K is used as a starting value for omega in the GallucciQuinn and Mooij parameterizations. The result of log(2) divided by the starting value for K is used as the starting value for t50 in the Weisberg parameterization.

If meth0="yngAge", then a starting value for t0 or L0 is found by algebraically solving the typical or original paramaterization, respectively, for t0 or L0 using the mean length of the first age with more than one data point as a "known" quantity. If meth0="poly" then a second-degree polynomial model is fit to the mean length-at-age data. The t0 starting value is set equal to the root of the polynomial that is closest to zero. The L0 starting value is set equal to the mean length at age-0 predicted from the polynomial function.

Starting values for the L1 and L3 parameters in the Schnute paramaterization and the L1, L2, and L3 parameters in the Francis parameterization may be found in two ways. If methEV="poly", then the starting values are the predicted length-at-age from a second-degree polynomial fit to the mean lengths-at-age data. If methEV="means" then the observed sample means at the corresponding ages are used. In the case where one of the supplied ages is fractional, then the value returned will be linearly interpolated between the mean lengths of the two closest ages. The ages to be used for L1

182 vbStarts

and L3 in the Schnute and Francis parameterizations are supplied as a numeric vector of length 2 in ages2use=. If ages2use=NULL then the minimum and maximum observed ages will be used. In the Francis method, L2 will correspond to the age half-way between the two ages in ages2use=. A warning will be given if L2<L1 for the Schnute method or if L2<L1 or L3<L2 for the Francis method.

Starting values can also be obtained by plotting the data with the model superimposed but tied to slider bars that allow the parameters to be interactively changed. One can change the parameters until a reasonable fit is observed and then use those valeus as starting values. Such a plot is constructed by including dynamicPlot=TRUE. The initial parameters for the slider bars are the starting values constructed as described above. It should be noted that the dynamic plot may show an error of "[tcl] can't get device image", but the plot will correctly update if the slider bar is adjusted.

Value

A list that contains reasonable starting values. Note that the parameters will be listed in the same order and with the same names as listed in vbFuns.

IFAR Chapter

12-Individual Growth.

Note

The 'original' and 'vonBertalanffy' and the 'typical' and 'BevertonHolt' parameterizations are synonymous.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL. See references in vbFuns.

See Also

See vbModels for a list of models and parameterizations used in **FSA** and vbFuns for functions that represent the von Bertalanffy parameterizations.

Examples

```
## Examples
data(SpotVA1)
vbStarts(tl~age,data=SpotVA1)
vbStarts(tl~age,data=SpotVA1,type="Original")
vbStarts(tl~age,data=SpotVA1,type="GQ")
vbStarts(tl~age,data=SpotVA1,type="Mooij")
vbStarts(tl~age,data=SpotVA1,type="Weisberg")
vbStarts(tl~age,data=SpotVA1,type="Francis",ages2use=c(0,5))
```

WhitefishLC 183

```
vbStarts(tl~age,data=SpotVA1,type="Schnute",ages2use=c(0,5))
vbStarts(tl~age,data=SpotVA1,type="Somers")
vbStarts(tl~age,data=SpotVA1,type="Somers2")
## Using a different method to find t0 and L0
vbStarts(tl~age,data=SpotVA1,meth0="yngAge")
vbStarts(tl~age,data=SpotVA1,type="original",meth0="yngAge")
## Using a different method to find the L1, L2, and L3
vbStarts(tl~age,data=SpotVA1,type="Francis",ages2use=c(0,5),methEV="means")
vbStarts(tl~age,data=SpotVA1,type="Schnute",ages2use=c(0,5),methEV="means")
## Example with a Plot
vbStarts(tl~age,data=SpotVA1,plot=TRUE)
vbStarts(tl~age,data=SpotVA1,type="original",plot=TRUE)
vbStarts(tl~age,data=SpotVA1,type="GQ",plot=TRUE)
vbStarts(tl~age,data=SpotVA1,type="Mooij",plot=TRUE)
vbStarts(tl~age,data=SpotVA1,type="Weisberg",plot=TRUE)
vbStarts(tl~age,data=SpotVA1,type="Francis",ages2use=c(0,5),plot=TRUE)
vbStarts(tl~age,data=SpotVA1,type="Schnute",ages2use=c(0,5),plot=TRUE)
vbStarts(tl~age,data=SpotVA1,type="Somers",plot=TRUE)
vbStarts(tl~age,data=SpotVA1,type="Somers2",plot=TRUE)
## Dynamic Plots Method -- ONLY RUN IN INTERACTIVE MODE
if (interactive()) {
 vbStarts(tl~age,data=SpotVA1,dynamicPlot=TRUE)
 vbStarts(tl~age,data=SpotVA1,type="original",dynamicPlot=TRUE)
 vbStarts(tl~age,data=SpotVA1,type="GQ",dynamicPlot=TRUE)\\
 vbStarts(tl~age,data=SpotVA1,type="Mooij",dynamicPlot=TRUE)
 vbStarts(tl~age,data=SpotVA1,type="Weisberg",dynamicPlot=TRUE)\\
 vbStarts(tl~age,data=SpotVA1,type="Francis",ages2use=c(0,5),dynamicPlot=TRUE)
  vbStarts(tl~age,data=SpotVA1,type="Schnute",ages2use=c(0,5),dynamicPlot=TRUE)
}
## See examples in vbFuns() for use of vbStarts() when fitting Von B models
```

WhitefishLC

Assessed ages from two readers on three structures for Lake Whitefish from Lake Champlain.

Description

Assessed ages from two readers on three structures for Lake Whitefish (*Coregonus clupeaformis*) from Lake Champlain in 2009.

Format

A data frame with 151 observations on the following 11 variables:

fishID A unique fish identification number.

184 WR79

tl Total length (in mm).

scale1 Assessed age from scales by first reader.

scale2 Assessed age from scales by second reader.

scaleC Consensus age from scales by both reader.

finray1 Assessed age from fin rays by first reader.

finray2 Assessed age from fin rays by second reader.

finrayC Consensus age from fin rays by both reader.

otolith1 Assessed age from otoliths by first reader.

otolith2 Assessed age from otoliths by second reader.

otolithC Consensus age from otoliths by both reader.

Topic(s)

- Age
- Ageing
- Precision
- Bias

Source

Data from Herbst, S.J. and J.E. Marsden. 2011. Comparison of precision and bias of scale, fin ray, and otolith age estimates for lake whitefish (*Coregonus clupeaformis*) in Lake Champlain. Journal of Great Lakes Research. 37:386-389. Contributed by Seth Herbst. **Do not use for other than educational purposes without permission from the author.**

Examples

data(WhitefishLC)
str(WhitefishLC)
head(WhitefishLC)

WR79

Age and length for a hypothetical sample from Westerheim and Ricker (1979).

Description

Age and length for a hypothetical sample in Westerheim and Ricker (1979).

Format

A data frame of 2369 observations on the following 3 variables:

ID Unique fish identifiers.

len Length of an individual fish.

age Age of an individual fish.

wrAdd 185

Details

Age-length data in 5-cm increments taken exactly from Table 2A of the source which was a sample from a hypothetical poplation in which year-class strength varied in the ratio 2:1 and the rate of increase in length decreased with age. Actual lengths in each 5-cm interval were simulated with a uniform distribution. The aged fish in this file were randomly selected and an assessed age was assigned according to the information in Table 2A.

Topic(s)

· Age-length key

Source

Simulated from Table 2A in Westerheim, S.J. and W.E. Ricker. 1979. Bias in using age-length key to estimate age-frequency distributions. Journal of the Fisheries Research Board of Canada. 35:184-189.

Examples

```
data(WR79)
str(WR79)
head(WR79)

## Extract the aged sample
WR79.aged <- subset(WR79,!is.na(age))
str(WR79.aged)

## Extract the length sample
WR79.length <- subset(WR79,is.na(age))
str(WR79.length)</pre>
```

wrAdd

Computes a vector of relative weights specific to a species in an entire data frame.

Description

This computes a vector that contains the relative weight specific to each species for all individuals in an entire data frame.

Usage

```
wrAdd(wt, ...)
## Default S3 method:
wrAdd(wt, len, spec, units = c("metric", "English"), ...)
## S3 method for class 'formula'
wrAdd(wt, data, units = c("metric", "English"), ...)
```

186 wrAdd

Arguments

wt	A numeric vector that contains weight measurements or a formula of the form wt~len+spec where "wt" generically represents the weight variable, "len" generically represents the length variable, and "spec" generically represents the species variable. Note that this formula can only contain three variables and they must be in the order of weight first, length second, species third.
len	A numeric vector that contains length measurements. Not used if wt is a formula.
spec	A character or factor vector that contains the specicas names. Not used if wt is a formula.
units	A string that indicates whether the weight and length data in formula are in ("metric" (DEFAULT; mm and g) or "English" (in and lbs) units.
data	A data frame that minimally contains variables of the the observed lengths, observed weights, and the species names given in the formula=.
	Not used.

Details

This computes a vector that contains the relative weight specific to each species for all individuals in an entire data frame. The vector can be appended to an existing data.frame to create a variable that contains the relative weights for each individual. The relative weight value will be NA for each individual for which a standard weight equation does not exist in WSlit, a standard weight equation for the units given in units= does not exist in WSlit, a standard weight equation for the 75th percentile does not exist in WSlit, or if the individual is shorter or longer than the lengths for which the standard weight equation should be applied. Either the linear or quadratic equation has been listed as preferred for each species, so only that equation will be used. The use of the 75th percentile is by far the most common and, because this function is designed for use on entire data frames, it will be the only percentile allowed. Thus, to use equations for other percentiles, one will have to use "manual" methods. See WSlit and wsVal for more details about types of equations, percentiles, finding which species have published standard weight equations, etc. See the examples for one method for changing species names to something that this function will recognize.

Value

Returns A numeric vector that contains the computed relative weights, in the same order as in data=.

IFAR Chapter

8-Condition.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

WSlit 187

See Also

See wsVal, WSlit, and psdAdd for related functionality. See mapvalues for help in changing species names to match those in WSlit.

Examples

```
## Create random data for three species
# just to control the randomization
set.seed(345234534)
dbt <- data.frame(species=factor(rep(c("Bluefin Tuna"),30)),tl=round(rnorm(30,1900,300),0))</pre>
dbt$wt <- round(4.5e-05*dbt$tl^2.8+rnorm(30,0,6000),1)
dbg <- data.frame(species=factor(rep(c("Bluegill"),30)),tl=round(rnorm(30,130,50),0))</pre>
dbg$wt <- round(4.23e-06*dbg$tl^3.316+rnorm(30,0,10),1)</pre>
dlb <- data.frame(species=factor(rep(c("Largemouth Bass"),30)),tl=round(rnorm(30,350,60),0))</pre>
dlb$wt <- round(2.96e-06*dlb$tl^3.273+rnorm(30,0,60),1)
df <- rbind(dbt,dbg,dlb)</pre>
str(df)
df$Wr1 <- wrAdd(wt~tl+species,data=df)</pre>
## same but with non-formula interface
df$Wr2 <- wrAdd(df$wt,df$tl,df$species)</pre>
## Same as above but using dplyr
if (require(dplyr)) {
  df <- mutate(df,Wr3a=wrAdd(wt,tl,species))</pre>
df
## Example with only one species in the data.frame
bg <- Subset(df,species=="Bluegill")</pre>
bg$Wr4 <- wrAdd(wt~tl+species,data=bg)</pre>
```

WSlit

All known standard weight equations.

Description

Parameters for all known standard weight equations.

Format

A data frame with observations on the following 13 variables:

species Species name.

units Units of measurements. Metric uses lengths in mm and weight in grams. English uses lengths in inches and weight in pounds.

type Type of equation (linear or quadratic).

WSlit

ref Reference quartile (75, 50, or 25).

measure The type of length measurement used – total length (TL) or fork length (FL).

method The type of method used to derive the equation (RLP,EmP, or Other).

min.len Minimum total length (mm or in, depending on units) for which the equation should be applied.

max.len Maximum total length (mm or in, depending on units) for which the equation should be applied.

int The intercept for the model.

slope The slope for the linear models or the linear coefficient for the quadratic equation.

quad The quadratic coefficient in the quadratic equations.

source Source of the equation. These match the sources given in Neumann et al. 2012.

comment Comments about use of equation.

Details

The minimum TL for the English units were derived by rounding the converted minimum TL for the metric units to what seemed like common units (inches, half inches, or quarter inches).

Topic(s)

- · Relative weight
- · Standard weight
- Condition

IFAR Chapter

8-Condition.

Source

Most of these equations can be found in Neumann, R.M., C.S. Guy, and D.W. Willis. 2012. Length, Weight, and Associated Indices. Chapter 14 in Zale, A.V., D.L. Parrish, and T.M. Sutton, editors. Fisheries Techniques. American Fisheries Society, Bethesda, MD.

Some species were not in Neumann et al (2012) and are noted as such in the comments variable.

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

See Also

See wsVal and wrAdd for related functionality.

Examples

```
data(WSlit)
str(WSlit)
head(WSlit)
```

wsVal

			-
WS	v	а	

Finds standard weight equation coefficients for a particular species.

Description

Returns a vector that contains all known or a subset of information about the standard weight equation for a given species, type of measurement units, and reference percentile.

Usage

```
wsVal(species = "List", units = c("metric", "English"), ref = 75,
simplify = FALSE)
```

Arguments

species	A string that contains the species name for which to find coefficients. See details.
units	A string that indicates whether the coefficients for the standard weight equation to be returned are in ("metric" (DEFAULT; mm and g) or "English" (in and lbs) units.
ref	A numeric that indicates which percentile the equation should be returned for. Note that the vast majority of equations only exist for the 75th percentile (DE-FAULT).
simplify	A logical that indicates whether the 'units', 'ref', 'measure', 'method', 'comments', and 'source' fields should be included (=FALSE) or not (=TRUE; DE-FAULT). See details.

Details

This function extract all known information from WSlit about the following standard weight equation,

$$log_{10}(Ws) = log_{10}(a) + blog_{10}(L) + blog_{10}(L)^2$$

See WSlit for more information about the meaning of each value returned.

Note from above that the coefficients are returned for the TRANSFORMED model. Thus, to obtain the standard weight (Ws), the returned coefficients are used to compute the common log of Ws which must then bed raised to the power of 10 to compute the Ws.

Value

A one row data frame from WSlit that contains all known information about the standard weight equation for a given species, type of measurement units, and reference percentile if simplify=FALSE. If simplify=TRUE then only the species; minimum and maximum length for which the standard equation should be applied; and intercept, slope, and quadratic coefficients for the standard weight equation. Note that the maximum length and the quadratic coefficient will not be returned if they do not exist in WSlit.

190 wsVal

If no arguments are given to this function, a species name is mis-spelled, or if a standard weight equation does not exist (in WSlit) for a particular species, then a warning will be issued and a list of species names will be printed.

IFAR Chapter

8-Condition.

Author(s)

Derek H. Ogle, <derek@derekogle.com>

References

Ogle, D.H. 2015. Introductory Fisheries Analyses with R. Chapman & Hall/CRC, Boca Raton, FL.

See Also

See wrAdd and WSlit for related functionality.

Examples

```
wsVal()
wsVal("Bluegill", units="metric")
wsVal("Bluegill", units="English")
wsVal("Bluegill", units="English", simplify=TRUE)
wsVal("Ruffe", units="metric", simplify=TRUE)
wsVal("Ruffe", units="metric", ref=50, simplify=TRUE)
```

Index

_	
*Topic datasets	confint.nlsBoot, 57
BluegillJL, 29	extraTests, 70
BrookTroutTH, 32	hoCoef, 95
ChinookArg, 50	hyperCI, 96
CodNorwegian, 51	ksTest, 104
CutthroatAL, 59	poiCI, 135
Ecoli, 66	*Topic manip
Mirex, 117	addRadCap, 3
PikeNY, 131	addZeroCatch, 5
PikeNYPartial1, 132	ageBias, 7
PSDlit, 144	ageKey, 13
SMBassLS, 163	agePrecision, 18
SMBassWB, 164	alkAgeDist, 21
SpotVA1, 165	alkMeanVar, 23
WhitefishLC, 183	bcFuns, 25
WR79, 184	capHistConvert,33
WSlit, 187	capHistSum, 40
*Topic hplot	catchCurve, 43
catchCurve, 43	chapmanRobson, 46
depletion, 60	chooseColors, 50
fitPlot, 74	col2rgbt, 52
growthModels,84	depletion, 60
hist.formula,92	expandCounts, 67
plotBinResp, 133	fact2num, 73
psdCalc, 139	fsaNews, 81
psdCI, 142	FSAUtils, 81
psdPlot, 145	gConvert, 82
residPlot, 156	growthModels,84
sumTable, 176	headtail, 91
tictactoe, 177	jolly, <mark>97</mark>
*Topic htest	kCounts, 101
ageBias,7	lagratio, 105
agePrecision, 18	lencat, 106
binCI, 27	lenFreqExpand, 111
bootstrap, 30	logbtcf, 113
catchCurve, 43	1wCompPreds, 114
chapmanRobson, 46	Mmethods, 118
compIntercepts, 53	mrClosed, 122
compSlopes, 55	oddeven, 129

psdAdd, 136	binconf, 28
psdVal, 148	binom.conf.int,28
removal, 151	binom.test, 28
Schnute, 161	BlueCrab, 62
se, 162	BluegillJL, 29
srStarts, 166	bootCase, <i>30</i> , <i>32</i>
stockRecruitment, 169	bootstrap, 30
validn, 179	BrookTroutTH, 32
vbStarts, 180	by, <i>174</i>
wrAdd, 185	•
wsVal, 189	capFirst(FSAUtils),81
*Topic misc	capHistConvert, 33, 40-42, 60
fishR, 74	capHistSum, 33, 34, 37, 40, 98, 100, 125, 127
perc, 130	catchCurve, 43, 49
rcumsum, 150	cdplot, <i>134</i>
Subset, 171	chapmanRobson, 45, 46
Summarize, 173	ChinookArg, 50
*Topic models	chooseColors, 16,50
fitPlot, 74	cm.colors, 51
plotBinResp, 133	CodNorwegian, 51
residPlot, 156	<pre>coef.catchCurve(catchCurve), 43</pre>
*Topic plot	coef.depletion(depletion), 60
ageKeyPlot, 16	col2rgb, <i>52</i>
3 3 4 4 5	col2rgbt, 52
abline, 79	colorRampPalette, 51
addRadCap, 3, 83	colors, <i>51</i> , <i>52</i>
addZeroCatch, 5	compare2, <i>11</i> , <i>12</i>
ageBias, 7, 19, 20	compIntercepts, 53, 56
ageKey, 13	compSlopes, 55
ageKeyPlot, 16	<pre>confint.bootCase (bootstrap), 30</pre>
agePrecision, 12, 18	<pre>confint.catchCurve(catchCurve), 43</pre>
agesurv, 45, 49	confint.chapmanRobson(chapmanRobson),
agesurvcl, 45, 49	46
aggregate, 174	confint.depletion(depletion), 60
AlewifeLH, <i>11</i> , <i>20</i>	confint.mrClosed1 (mrClosed), 122
alkAgeDist, <i>14</i> , <i>15</i> , 21, 25	confint.mrClosed2 (mrClosed), 122
alkIndivAge, 18, 23-25	confint.mrOpen(jolly), 97
alkIndivAge (ageKey), 13	confint.nlsBoot, 57
alkMeanVar, 14, 15, 23	<pre>confint.nlsboot(confint.nlsBoot), 57</pre>
alkPlot, 15	confint.removal(removal), 151
alkPlot (ageKeyPlot), 16	cumsum, <i>150</i>
alkprop, 22, 23	CutthroatAL, 59, 99
anova, 71	CutthroatALf, 60
anova.catchCurve (catchCurve), 43	
anova.depletion (depletion), 60	darter, 62
	deplet, <i>62</i> , <i>63</i>
basicStats, <i>175</i>	depletion, 60, <i>154</i>
bcFuns, 25	describe, 175
binCI, 27, 125, 127, 142	descriptive, 42

diff, 105	kCounts, 101
drop.levels, 172	knit, <i>103</i>
droplevels, <i>6</i> , <i>172</i>	kPvalue (kCounts), 101
dunn.test, 65, 66	kruskal, <i>65</i> , <i>66</i>
dunnTest, 64	kruskal.test,66
	kruskalmc, <i>65</i> , <i>66</i>
Ecoli, 66	ks.test, <i>104</i> , <i>105</i>
error.bars, 79	ksTest, 104
expandCounts, 67, 112	
expandLenFreq, 68	lagratio, 105
expandLenFreq(lenFreqExpand), 111	lencat, 106, 111, 112, 140, 143, 147, 149
extraSS (extraTests), 70	lenFregExpand, 111
extraTests, 70	lineplot.CI, 77, 79
	LobsterPEI, 62
fact2num, 73	logbtcf, 113
filter, <i>172</i>	logisticFuns, 162
filterD, 6	logisticFuns (growthModels), 84
filterD (Subset), 171	logisticModels (growthModels), 84
fishR, 74	lrt (extraTests), 70
fitPlot, 74, 134, 159	1rtest, 71
formatC, <i>103</i>	lwCompPreds, 114
FSA, 80	1110011p1 1 000, 11 1
FSA-package (FSA), 80	M.empirical, <i>120</i> , <i>121</i>
FSANews (fsaNews), 81	mapvalues, 117, 117, 138, 187
fsaNews, 81	metaM, 45, 49
FSAUtils, 81	metaM (Mmethods), 118
13/00113, 01	Mirex, 117
gConvert, 4, 82	Mmethods, 118
GompertzFuns, 162	mrClosed, 37, 40–42, 100, 122
GompertzFuns (growthModels), 84	mrN. single, <i>127</i>
GompertzModels (growthModels), 84	mrOpen, 37, 40–42, 60, 127
growthModels, 84	mrOpen (jolly), 97
gi oweiiilode13, 04	multhist, 94
headtail, 91	marchise, 74
heat.colors, 51	nlsBoot, 57
hist, 94	nlsResiduals, <i>159</i>
hist.bootCase (bootstrap), 30	nobs, 180
hist.formula, 92	110003, 1000
hoCoef, 95	oddeven, 129
htest (confint.nlsBoot), 57	outlierTest, 159
htest.bootCase (bootstrap), 30	0002101.1000, 100
htest.nlsBoot, 96	p.adjust,55
hyperCI, 96, 125, 127	p.adjust.methods, 55
Hyper C1, 70, 123, 127	p.adjustment.methods, 64, 65
is.even (oddeven), 129	pairw.kw, 65
is.na, 180	palette, 52
is.odd (oddeven), 129	paletteChoices (chooseColors), 50
15.044 (0446 (011), 12)	pcumsum (rcumsum), 150
jolly, 97	perc, 130
	r -: -,

PikeNY, 131	rstudent, 159
PikeNYPartial1, 132	
plot.ageBias (ageBias), 7	schnabel, 127
plot.agePrec (agePrecision), 18	Schnute, 88, 161
plot.bootCase (bootstrap), 30	SchnuteModels (Schnute), 161
plot.CapHist(capHistSum), 40	se, 162, <i>163</i>
plot.catchCurve (catchCurve), 43	SMBassLS, <i>62</i> , 163
plot.chapmanRobson(chapmanRobson), 46	SMBassWB, 164
plot.depletion (depletion), 60	SnapperHG2, 22
plot.mrClosed2 (mrClosed), 122	SpotVA1, 165
plotBinResp, 133	SpotVA2, <i>166</i>
plotCI, 9, 177, 178	srFuns, <i>167, 168</i>
plotmath, 85, 161, 170	<pre>srFuns (stockRecruitment), 169</pre>
plotmeans, 79	srModels, 168
poiCI, <i>125–127</i> , 135	<pre>srModels(stockRecruitment), 169</pre>
pois.exact, 136	srStarts, 166, <i>170</i>
posthoc.kruskal.nemenyi.test, 65, 66	stockRecruitment, 169
<pre>predict.bootCase (bootstrap), 30</pre>	Subset, 171
<pre>predict.nlsBoot(confint.nlsBoot), 57</pre>	subset, <i>172</i>
<pre>print.compIntercepts (compIntercepts),</pre>	Summarize, 173, <i>177</i>
53	summary, <i>173-175</i>
<pre>print.compSlopes (compSlopes), 55</pre>	<pre>summary.ageBias(ageBias), 7</pre>
print.dunnTest (dunnTest), 64	<pre>summary.agePrec (agePrecision), 18</pre>
print.extraTest (extraTests), 70	<pre>summary.catchCurve(catchCurve), 43</pre>
print.metaM (Mmethods), 118	summary.chapmanRobson(chapmanRobson),
psdAdd, 136, 140, 143, 145, 147, 149, 187	46
psdCalc, 138, 139, 142, 145, 147, 149, 178	summary.depletion (depletion), 60
psdCI, 139, 140, 142	summary.mrClosed1 (mrClosed), 122
PSDlit, 137, 138, 140, 143, 144, 147, 149	summary.mrClosed2 (mrClosed), 122
psdPlot, 138, 140, 143, 145, 145, 149	summary.mrOpen(jolly), 97
psdVal, 138–140, 143, 145, 147, 148, 178	summary.nlsBoot, 58
purl, <i>103</i>	summary.removal(removal), 151
purl2 (kCounts), 101	summaryBy, <i>175</i>
[(((((((((((((((((((sumTable, 176
rainbow, 51	SunfishIN, 127
Rcapture, 34, 36, 37	,
rcumsum, 140, 143, 147, 149, 150	table, <i>173–175</i>
regLine, 79	tapply, <i>174–177</i>
removal, 63, 151	terrain.colors, 51
reproInfo(kCounts), 101	tictactoe, 140, 143, 145, 147, 149, 177
residPlot, 79, 156	topo.colors, 51
residualPlots, 159	TukeyHSD, <i>54</i>
rgb, 52	•
rich.colors, 51	valid.n, <i>180</i>
RichardsFuns, 162	validn, 179
RichardsFuns (growthModels), 84	vbFuns, 162, 182
RichardsModels (growthModels), 84	vbFuns (growthModels), 84
RMark, <i>35–37</i>	vbModels, 182
rstandard, 159	vbModels (growthModels), 84

```
vbStarts, 180

wblake2, 165

WhitefishLC, 20, 183

WR79, 184

wrAdd, 138, 185, 188, 190

WSlit, 186, 187, 187, 189, 190

wsVal, 186–188, 189

xtabs, 174, 175
```