

Package ‘gbm.auto’

July 22, 2025

Title Automated Boosted Regression Tree Modelling and Mapping Suite

Version 2024.10.01

Description Automates delta log-normal boosted regression tree abundance prediction. Loops through parameters provided (LR (learning rate), TC (tree complexity), BF (bag fraction)), chooses best, simplifies, & generates line, dot & bar plots, & outputs these & predictions & a report, makes predicted abundance maps, and Unrepresentativeness surfaces. Package core built around 'gbm' (gradient boosting machine) functions in 'dismo' (Hijmans, Phillips, Leathwick & Jane Elith, 2020 & ongoing), itself built around 'gbm' (Greenwell, Boehmke, Cunningham & Metcalfe, 2020 & ongoing, originally by Ridgeway). Indebted to Elith/Leathwick/Hastie 2008 'Working Guide' <[doi:10.1111/j.1365-2656.2008.01390.x](https://doi.org/10.1111/j.1365-2656.2008.01390.x)>; workflow follows Appendix S3. See <<https://www.simondedman.com/>> for published guides and papers using this package.

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Depends R (>= 3.5.0)

Imports beapr (>= 1.2), dismo (>= 1.3-14), dplyr (>= 1.0.9), gbm (>= 2.1.1), ggmap (>= 3.0.2), ggplot2 (>= 3.4.2), ggspatial (>= 1.1.9), lifecycle, lubridate (>= 1.9.2), mapplots (>= 1.5), Metrics (>= 0.1.4), readr (>= 2.1.4), sf (>= 0.9-7), stars (>= 0.6-3), starsExtra (>= 0.2.7), stats (>= 3.3.1), stringi (>= 1.6.1), tidyselect (>= 1.2.0), viridis (>= 0.6.4)

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Adult_Females	<i>Data: Numbers of 4 adult female rays caught in 2137 Irish Sea trawls, 1994 to 2014</i>
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Description

2137 capture events of adult female cuckoo, thornback, spotted and blonde rays in the Irish Sea from 1994 to 2014 by the ICES IBTS, including explanatory variables: Length Per Unit Effort in that area by the commercial fishery, depth, temperature, distance to shore, and current speed at the bottom.

Usage

data(Adult_Females)

Format

A data frame with 2137 rows and 13 variables:

Longitude Decimal longitudes in the Irish Sea

Latitude Decimal latitudes in the Irish Sea

Haul_Index ICES IBTS area, survey, station, and year

F_LPUE Commercial fishery LPUE in Kg/Hr

Depth Metres, decimal

Temperature Degrees, decimal

Salinity PPM

Distance_to_Shore Metres, decimal

Current_Speed Metres per second at the seabed

Cuckoo Numbers of cuckoo rays caught, standardised to 1 hour

Thornback Numbers of thornback rays caught, standardised to 1 hour

Blonde Numbers of blonde rays caught, standardised to 1 hour

Spotted Numbers of spotted rays caught, standardised to 1 hour

Author(s)

Simon Dedman, <simondedman@gmail.com>

Source

<http://datras.ices.dk>

AllPreds_E

Data: Predicted abundances of 4 ray species generated using gbm.auto

Description

Predicted abundances of 4 ray species generated using gbm.auto, and Irish commercial beam trawler effort 2012.

Usage

```
data(AllPreds_E)
```

Format

A data frame with 378570 rows and 7 variables:

Latitude Decimal latitudes in the Irish Sea

Longitude Decimal longitudes in the Irish Sea

Cuckoo Predicted abundances of cuckoo rays in the Irish Sea, generated using gbm.auto

Thornback Predicted abundances of thornback rays in the Irish Sea, generated using gbm.auto

Blonde Predicted abundances of blonde rays in the Irish Sea, generated using gbm.auto

Spotted Predicted abundances of spotted rays in the Irish Sea, generated using gbm.auto

Effort Irish commercial beam trawler effort 2012

Author(s)

Simon Dedman, <simondedman@gmail.com>

AllScaledData

Data: Scaled abundance data for 2 subsets of 4 rays in the Irish Sea, by gbm.cons

Description

A dataset containing the output of the gbm.cons example run, conservation priority areas within the Irish Sea for juvenile and adult female cuckoo, blonde, thornback and spotted rays.

Usage

```
data(AllScaledData)
```

Format

A data frame with 378570 rows and 3 variables:

Longitude Decimal longitudes in the Irish Sea

Latitude Decimal latitudes in the Irish Sea

allscaled Relative abundance. Each juvenile and adult female cuckoo, blonde, thornback and spotted ray scaled to 1 and added together

Author(s)

Simon Dedman, <simondedman@gmail.com>

breaks.grid*Defines breakpoints for draw.grid and legend.grid; mapplots fork*

Description

Defines breakpoints from values in `grd` with options to exclude outliers, set number of bins, and include a dedicated zero column. Forked by SD 05/01/2019 to add 'lo', else bins always begin at 0, killing plotting when all data are in a tight range at high values e.g. 600:610

Usage

```
breaks.grid(grd, quantile = 0.975, ncol = 12, zero = TRUE)
```

Arguments

<code>grd</code>	An array produced by <code>make.grid</code> or a list produced by <code>make.multigrid</code> or a vector of positive values.
<code>quantile</code>	The maximum value of the breaks will be determined by the quantile given here. This can be used to deal with outlying values in <code>grd</code> . If <code>quantile = 1</code> then the maximum value of the breaks will be the same as the maximum value in <code>grd</code> .
<code>ncol</code>	Number of colours to be used, always one more than the number of breakpoints. Defaults to 12.
<code>zero</code>	Logical, should zero be included as a separate category? Defaults to TRUE.

Value

A vector of breakpoints for `draw.grid` in `mapplots`

Author(s)

Simon Dedman, <simondedman@gmail.com>

Hans Gerritsen

Examples

```
breaks.grid(100,ncol=6)
breaks.grid(100,ncol=5,zero=FALSE)

# create breaks on the log scale
exp(breaks.grid(log(10000),ncol=4,zero=FALSE))
```

calibration

calibration

Description

Internal use only. Jane Elith/John Leathwick 17th March 2005. Calculates calibration statistics for either binomial or count data but the family argument must be specified for the latter a conditional test for the latter will catch most failures to specify the family.

Usage

```
calibration(obs, preds, family = c("binomial", "bernoulli", "poisson"))
```

Arguments

obs	Observed data.
preds	Predicted data.
family	Statistical distribution family. Choose one.

Value

roc & calibration stats internally within gbm runs e.g. in gbm.auto.

Author(s)

Simon Dedman, <simondedman@gmail.com>

gbm.auto

Automated Boosted Regression Tree modelling and mapping suite

Description

Automates delta log normal boosted regression trees abundance prediction. Loops through all permutations of parameters provided (learning rate, tree complexity, bag fraction), chooses the best, then simplifies it. Generates line, dot and bar plots, and outputs these and the predictions and a report of all variables used, statistics for tests, variable interactions, predictors used and dropped, etc. If selected, generates predicted abundance maps, and Unrepresentativeness surfaces. See www.GitHub.com/SimonDedman/gbm.auto for issues, feedback, and development suggestions. See SimonDedman.com for links to walkthrough paper, and papers and thesis published using this package.

Usage

```
gbm.auto(
  grids = NULL,
  samples,
  expvar,
  resvar,
  randomvar = FALSE,
  tc = c(2),
  lr = c(0.01, 0.005),
  bf = 0.5,
  offset = NULL,
  n.trees = 50,
  ZI = "CHECK",
  fam1 = c("bernoulli", "binomial", "poisson", "laplace", "gaussian"),
  fam2 = c("gaussian", "bernoulli", "binomial", "poisson", "laplace"),
  simp = TRUE,
  gridslat = 2,
  gridslon = 1,
  samplesGridsAreaScaleFactor = 1,
  multiplot = TRUE,
  cols = grey.colors(1, 1, 1),
  linesfiles = TRUE,
  smooth = FALSE,
  savedir = tempdir(),
  savegbm = TRUE,
  loadgbm = NULL,
  varint = TRUE,
  map = TRUE,
  shape = NULL,
  RSB = TRUE,
  BnW = TRUE,
  alerts = TRUE,
  pngtype = c("cairo-png", "quartz", "Xlib"),
  gaus = TRUE,
  MLEvaluate = TRUE,
  brv = NULL,
  grv = NULL,
  Bin_Preds = NULL,
  Gaus_Preds = NULL,
  ...
)
```

Arguments

grids	Explanatory data to predict to. Import with (e.g.) read.csv and specify object name. Defaults to NULL (won't predict to grids).
samples	Explanatory and response variables to predict from. Keep col names short (~17 characters max), no odd characters, spaces, starting numerals or terminal peri-

	ods. Spaces may be converted to periods in directory names, underscores won't. Can be a subset of a large dataset.
expvar	Vector of names or column numbers of explanatory variables in 'samples': c(1,3,6) or c("Temp","Sal"). No default.
resvar	Name or column number(s) of response variable in samples: 12, c(1,4), "Rock-fish". No default. Column name is ideally species name.
randomvar	Add a random variable (uniform distribution, 0-1) to the expvars, to see whether other expvars perform better or worse than random.
tc	Permutations of tree complexity allowed, can be vector with the largest sized number no larger than the number of explanatory variables e.g. c(2,7), or a list of 2 single numbers or vectors, the first to be passed to the binary BRT, the second to the Gaussian, e.g. tc = list(c(2,6), 2) or list(6, c(2,6)).
lr	Permutations of learning rate allowed. Can be a vector or a list of 2 single numbers or vectors, the first to be passed to the binary BRT, the second to the Gaussian, e.g. lr = list(c(0.01,0.02),0.0001) or list(0.01,c(0.001, 0.0005)).
bf	Permutations of bag fraction allowed, can be single number, vector or list, per tc and lr. Defaults to 0.5.
offset	Column number or quoted name in samples, containing offset values relating to the samples. A numeric vector of length equal to the number of cases. Similar to weighting, see https://towardsdatascience.com/offsetting-the-model-logic-to-implementation-7e333bc25798 .
n.trees	From gbm.step, number of initial trees to fit. Can be single or list but not vector i.e. list(fam1,fam2).
ZI	Are data zero-inflated? TRUE FALSE "CHECK". Choose one. TRUE: delta BRT, log-normalised Gaus, reverse log-norm and bias corrected. FALSE: do Gaussian only, no log-normalisation. "CHECK": Tests data for you. Default is "CHECK". TRUE and FALSE aren't in quotes, "CHECK" is.
fam1	Probability distribution family for 1st part of delta process, defaults to "bernoulli". Choose one.
fam2	Probability distribution family for 2nd part of delta process, defaults to "gaussian". Choose one.
simp	Try simplifying best BRTs?
gridslat	Column number for latitude in 'grids'.
gridslon	Column number for longitude in 'grids'.
samplesGridsAreaScaleFactor	Scale up or down factor so values in the predict-to pixels of 'grids' match the spatial scale sampled by rows in 'samples'. Default 1 means no change.
multiplot	Create matrix plot of all line files? Default true. turn off if big n of exp vars causes an error due to margin size problems.
cols	Barplot colour vector. Assignment in order of explanatory variables. Default <i>1white: white bars black borders</i> . 'I' repeats.
linesfiles	Save individual line plots' data as csv's? Default TRUE.
smooth	Apply a smoother to the line plots? Default FALSE.

savedir	Save outputs to a temporary directory (default) else change to current directory e.g. "/home/me/folder". Do not use getwd() here.
savegbm	Save gbm objects and make available in environment after running? Open with load("Bin_Best_Model") Default TRUE.
loadgbm	Relative or (very much preferably) absolute location of folder containing Bin_Best_Model and Gaus_Best_Model. If set will skip BRT calculations and do predicted maps and csvs. Avoids re-running BRT models again (the slow bit), can run normally once with savegbm=T then multiple times with new grids & loadgbm to predict to multiple grids e.g. different seasons, areas, etc. Default NULL, character vector, "/" for working directory.
varint	Calculate variable interactions? Default:TRUE, FALSE for error: "contrasts can be applied only to factors with 2 or more levels".
map	Save abundance map png files?
shape	Enter the full path to downloaded map e.g. coastline shapefile, possibly from gbm.basemap, typically Crop_Map.shp, including the .shp. Can also name an existing object in the environment, read in with sf::st_read. Default NULL, in which case bounds calculated by gbm.mapsf which then calls gbm.basemap to download and auto-generate the base map.
RSB	Run Unrepresentativeness surface builder? Default TRUE.
BnW	Repeat maps in black and white e.g. for print journals. Default TRUE.
alerts	Play sounds to mark progress steps. Default TRUE but running multiple small BRTs in a row (e.g. gbm.loop) can cause RStudio to crash.
pngtype	Filetype for png files, alternatively try "quartz" on Mac. Choose one.
gaus	Do family2 (typically Gaussian) runs as well as family1 (typically Bin)? Default TRUE.
MLEvaluate	do machine learning evaluation metrics & plots? Default TRUE.
brv	Dummy param for package testing for CRAN, ignore.
grv	Dummy param for package testing for CRAN, ignore.
Bin_Preds	Dummy param for package testing for CRAN, ignore.
Gaus_Preds	Dummy param for package testing for CRAN, ignore.
...	Optional arguments for gbm.step (dismo package) arguments n.trees and max.trees, both of which can be added in list(1,2) format to pass to fam1 and 2; for gbm.mapsf colourscale, heatcolours, colournumber, and others.

Details

Errors and their origins:

1. install ERROR: dependencies 'rgdal', 'rgeos' are not available for package 'gbm.auto'. For Linux/*buntu systems, in terminal, type: 'sudo apt install libgeos-dev', 'sudo apt install libproj-dev', 'sudo apt install libgdal-dev'.
2. Error in FUN(X[[i]], ...) : only defined on a data frame with all numeric variables. Check your variable types are correct, e.g. numerics haven't been imported as factors because there's an errant first row of text information before the data. Remove NA rows from the response variable if present: convert blank cells to NA on import with read.csv(x, na.strings = "") then `samples2 <- samples[-which(is.na(samples[,resvar_column_number]))]`

3. At BF=0.5, if nrows <= 42, gbm.step will crash. Use gbm.bfcheck to determine optimal viable BF size.
4. Maps/plots don't work/output. If on a Mac, try changing pngtype to "quartz".
5. Error in while (delta.deviance > tolerance.test & n.fitted < max.trees): missing value where TRUE/FALSE needed. If running a zero-inflated delta model (bernoulli/bin & gaussian/gaus), Data are expected to contain zeroes (lots of them in zero-inflated cases), have you already filtered them out, i.e. are only testing the positive cases? Or do you only have positive cases? If so only run (e.g.) Gaussian: set ZI to FALSE.
6. Error in round(gbm.object\$cv.statistics\$deviance.mean, 4) : non-numeric argument to mathematical function. LR or BF probably too low in earlier BRT (normally Gaus run with highest TC).
7. Error in if (n.trees > x\$n.trees) argument is of length zero. LR or BF probably too low in earlier BRT (normally Gaus run with highest TC).
8. Error in gbm.fit(x, y, offset = offset, distribution = distribution, w = w): The dataset size is too small or subsampling rate is too large: nTrain*bag.fraction <= n.minobsinnode. LR or BF probably too low in earlier BRT (normally Gaus run with highest TC). It may be that you don't have enough positive samples to run BRT modelling. Run gbm.bfcheck to check recommended minimum BF size.
9. Warning message: In cor(y_i, u_i) : the standard deviation is zero. LR or BF probably too low in earlier BRT (normally Gaus run with highest TC). It may be that you don't have enough positive samples to run BRT modelling. Run gbm.bfcheck to check recommended minimum BF size. Similarly: glm.fit: fitted probabilities numerically 0 or 1 occurred, and glm.fit: algorithm did not converge. Similarly: Error in if (get(paste0("Gaus_BRT", ".tc", j, ".lr", k, ".bf", 1))\$self.statistics\$correlation[[1]]): argument is of length zero. See also: Error 15.
10. Anomalous values can obfuscate clarity in line plots e.g. salinity range 32:35ppm but dataset has errant 0 value: plot axis will be 0:35, and 99.99% of the data will be in the tiny bit at the right. Clean your data beforehand.
11. Error in plot.new() : figure margins too large: In RStudio, adjust plot pane (usually bottom right) to increase its size. Still fails? Set multiplot=FALSE.
12. Error in dev.print(file = paste0("./", names(samples[i]), "/pred_dev_bin.jpeg")): can only print from a screen device. An earlier failed run (e.g. LR/BF too low) left a plotting device open. Close it with: 'dev.off()'.
 - 13. RStudio crashed: set alerts=F and pause cloud sync programs if outputting to a synced folder.
14. Error in grDevices::dev.copy(device = function (filename = "Rplot%03d.jpeg", could not open file './resvar/pred_dev_bin.jpeg' (or similar). Your resvar column name contains an illegal character e.g. /&'_. Fix with colnames(samples)[n] <- "BetterName".
15. Error in gbm.fit: Poisson requires the response to be a positive integer. If running Poisson distributions, ensure the response variables are positive integers, but if they are, try a smaller LR.
16. If lineplots of factorial variables include empty columns be sure to remove unused levels with samples %<>% droplevels() before the gbm.auto run.
17. Error in seq.default(from = min(x\$var.levels[[i.var[i]]]), to = max(x\$var.levels[[i.var[i]]]): 'from' must be a finite number. If you logged any expvars with log() and they has zeroes in them, those zeroes became imaginary numbers. Use log1p() instead.

18. Error in loadNamespace... 'dismo' 1.3-9 is being loaded, but >= 1.3.10 is required: first do `remotes::install_github("rsatial/dismo")` then `library(dismo)`.
19. Error in `if (scope >= 160) res <- "c"` : missing value where TRUE/FALSE needed. Check `gridslat` and `gridslon` are indexing the correct columns in `grids`.

ALSO: check this section in the other functions run by `gbm.auto` e.g. `gbm.mapsf`, `gbm.basemap`. Use `traceback()` to find the source of errors.

I strongly recommend that you download papers 1 to 5 (or just the doctoral thesis) on <http://www.simondedman.com>, with emphasis on P4 (the guide) and P1 (statistical background). Elith et al 2008 (<https://besjournals.onlinelibrary.wiley.com/doi/10.1111/j.1365-2656.2008.01390.x>) is also strongly recommended. Just because you CAN try every conceivable combination of `tc`, `lr`, `bf`, `all`, at once doesn't mean you should. Try a range of `lr` in shrinking orders of magnitude from 0.1 to 0.000001, find the best, THEN try `tc` `c(2, n.expvars)`, find the best THEN `bf` `c(0.5, 0.75, 0.9)` and then in between if either outperform 0.5.

Value

Line, dot and bar plots, a report of all variables used, statistics for tests, variable interactions, predictors used and dropped, etc. If selected, generates predicted abundance maps, and Unrepresentativeness surface. Biggest Interactions in the report csv: see `?dismo::gbm.interactions`.

Author(s)

Simon Dedman, <simondedman@gmail.com>

Examples

```
# Not run. Note: grids file was heavily cropped for CRAN upload so output map
# predictions only cover patchy chunks of the Irish Sea, not the whole area.
# Full versions of these files:
# https://drive.google.com/file/d/1WHYpftP3roozVKwi_R_IpW7tLZIhZA7r
# /view?usp=sharing
library(gbm.auto)
data(grids)
data(samples)
# Set your working directory
gbm.auto(grids = grids, samples = samples, expvar = c(4:8, 10), resvar = 11,
tc = c(2,7), lr = c(0.005, 0.001), ZI = TRUE, savegbm = FALSE)
```

gbm.basemap

Creates Basemaps for Gbm.auto mapping from your data range

Description

Downloads unzips crops & saves NOAAs global coastline shapefiles to user-set box. Use for 'shape' in `gbm.map`. If downloading in RStudio uncheck "Use secure download method for HTTP" in Tools > Global Options > Packages. Simon Dedman, 2015/6 simondedman@gmail.com [GitHub.com/SimonDedman/gbm.auto](https://github.com/SimonDedman/gbm.auto)

Usage

```
gbm.basemap(
  bounds = NULL,
  grids = NULL,
  gridslat = NULL,
  gridslon = NULL,
  getzip = TRUE,
  zipvers = "2.3.7",
  savedir = tempdir(),
  savename = "Crop_Map",
  res = "CALC",
  extrabounds = FALSE
)
```

Arguments

bounds	Region to crop to: c(xmin,xmax,ymin,ymax).
grids	If bounds unspecified, name your grids database here.
gridslat	If bounds unspecified, specify which column in grids is latitude.
gridslon	If bounds unspecified, specify which column in grids is longitude.
getzip	Download & unpack GSHHS data to WD? "TRUE" else absolute/relative reference to GSHHS_shp folder, including that folder.
zipvers	GSHHS version, in case it updates. Please email developer (SD) if this is incorrect.
savedir	Save outputs to a temporary directory (default) else change to current directory e.g. "/home/me/folder". Do not use getwd() here.
savename	Shapefile save-name, no shp extension, default is "Crop_Map"
res	Resolution, 1:5 (low:high) OR c,l,i,h,f (coarse, low, intermediate, high, full) or "CALC" to calculate based on bounds. Choose one.
extrabounds	Grow bounds 16pct each direction to expand rectangular datasets basemaps over the entire square area created by basemap in mapplots.

Details

errors and their origins:

1. Error in setwd(getzip) : cannot change working directory If you've specified the location of the local GSHHS_shp folder, ensure you're in the correct directory relative to it. This error means it looked for the folder and couldn't find it.
2. subscript out of bounds: can't crop world map to your bounds. Check lat/lon are the right way around: check gridslat and gridslon point to the correct columns for lat and lon in grids, and those columns named (something like) lat and lon, ARE ACTUALLY the latitudes and longitudes, and not the wrong way around.
3. If your download is timing out use options(timeout = 240).
4. Error in if (scope >= 160) res <- "c" : missing value where TRUE/FALSE needed. Check gridslat and gridslon are indexing the correct columns in grids.

Value

basemap coastline file for gbm.map in gbm.auto. "cropshp" SpatialPolygonsDataFrame in local environment & user-named files in "CroppedMap" folder. Load later with maptools function: `MyMap <- sf::st_read(dsn = "/CroppedMap/Crop_Map.shp", layer = "Crop_Map", quiet = TRUE)`

Author(s)

Simon Dedman, <simondedman@gmail.com>

Examples

```
# Not run: downloads and saves external data.
data(samples)
mybounds <- c(range(samples[,3]),range(samples[,2]))
gbm.basemap(bounds = mybounds, getzip = "./GSHHS_shp/",
savename = "My_Crop_Map", res = "f")
# In this example GSHHS folder already downloaded to the working directory
# hence I pointed getzip at that rather than having it download the zip again
```

gbm.bfcheck

Calculates minimum Bag Fraction size for gbm.auto

Description

Provides minimum bag fractions for gbm.auto, preventing failure due to bf & samples rows limit.
Simon Dedman, 2016, simondedman@gmail.com, [GitHub.com/SimonDedman/gbm.auto](https://github.com/SimonDedman/gbm.auto)

Usage

```
gbm.bfcheck(samples, resvar, ZI = "CHECK", grv = NULL)
```

Arguments

<code>samples</code>	Samples dataset, same as gbm.auto.
<code>resvar</code>	Response variable column in samples.
<code>ZI</code>	Are samples zero-inflated? TRUE/FALSE/"CHECK".
<code>grv</code>	Dummy param for package testing for CRAN, ignore.

Value

Prints minimum Bag Fraction size for gbm.auto.

Author(s)

Simon Dedman, <simondedman@gmail.com>

Examples

```
data(samples)
gbm.bfcheck(samples = samples, resvar = "Cuckoo")
```

gbm.cons

Conservation Area Mapping

Description

Runs gbm.auto for multiple subsets of the same overall dataset and scales the combined results, leading to maps which highlight areas of high conservation importance for multiple species in the same study area e.g. using juvenile and adult female subsets to locate candidate nursery grounds and spawning areas respectively.

Usage

```
gbm.cons(
  mygrids,
  subsets,
  alerts = TRUE,
  map = TRUE,
  BnW = TRUE,
  resvars,
  gbmautos = TRUE,
  savedir = tempdir(),
  expvars,
  tcs = NULL,
  lrs = rep(list(c(0.01, 0.005)), length(resvars)),
  bfs = rep(0.5, length(resvars)),
  ZIs = rep("CHECK", length(resvars)),
  colss = rep(list(grey.colors(1, 1, 1)), length(resvars)),
  linesfiless = rep(FALSE, length(resvars)),
  savegbms = rep(TRUE, length(resvars)),
  varints = rep(TRUE, length(resvars)),
  maps = rep(TRUE, length(resvars)),
  RSBs = rep(TRUE, length(resvars)),
  BnWs = rep(TRUE, length(resvars)),
  zeroes = rep(TRUE, length(resvars)),
  shape = NULL,
  pngtype = c("cairo-png", "quartz", "Xlib"),
  gridslat = 2,
  gridslon = 1,
  grids = NULL
)
```

Arguments

mygrids	Gridded lat+long+data object to predict to.
subsets	Subset name(s): character; single or vector, corresponding to matching-named dataset objects e.g. read in by read.csv().
alerts	Play sounds to mark progress steps.
map	Produce maps.
BnW	Also produce B&W maps?
resvars	Vector of resvars cols from dataset objects for gbm.autos, length(subsets)*species, no default.
gbmautos	Do gbm.auto runs for species? Default TRUE, set FALSE if already run and output files in expected directories.
savedir	Save outputs to a temporary directory (default) else change to current directory e.g. "/home/me/folder". Do not use getwd() here.
expvars	List object of expvar vectors for gbm.autos, length = no. of subsets * no. of species. No default.
tcs	Gbm.auto parameters, auto-calculated below if not provided by user.
lrs	Gbm.auto parameter, uses defaults if not provided by user.
bfs	Gbm.auto parameter, uses defaults if not provided by user.
ZIs	Gbm.auto parameter, autocalculated below if not provided by user. Choose one entry.
colss	Gbm.auto parameter, uses defaults if not provided by user.
linesfiless	Gbm.auto parameter, uses defaults if not provided by user.
savegbms	Gbm.auto parameter, uses defaults if not provided by user.
varints	Gbm.auto parameter, uses defaults if not provided by user.
maps	Gbm.auto parameter, uses defaults if not provided by user.
RSBs	Gbm.auto parameter, uses defaults if not provided by user.
BnWs	Gbm.auto parameter, uses defaults if not provided by user.
zeroes	For breaks.grid, include zero-only category in colour breakpoints and subsequent legend. Defaults to TRUE.
shape	Coastline file for gbm.map.
pngtype	File-type for png files, alternatively try "quartz" on Mac. Choose one.
gridslat	Per Gbm.auto defaults to 2.
gridslon	Per Gbm.auto defaults to 1.
grids	Dummy param for package testing for CRAN, ignore.

Value

Maps via gbm.map & saved data as csv file.

Author(s)

Simon Dedman, <simondedman@gmail.com>

Examples

```
# Not run: downloads and saves external data.
data(grid)
gbm.cons(mygrids = grids, subsets = c("Juveniles", "Adult_Females"),
  resvars = c(44:47, 11:14),
  expvars = list(c(4:11, 15, 17, 21, 25, 29, 37),
    c(4:11, 15, 18, 22, 26, 30, 38),
    c(4:11, 15, 19, 23, 27, 31),
    c(4:11, 15, 20, 24, 28, 32, 39),
    4:10, 4:10, 4:10, 4:10),
  tcs = list(c(2, 14), c(2, 14), 13, c(2, 14), c(2, 6), c(2, 6), 6,
    c(2, 6)),
  lrs = list(c(0.01, 0.005), c(0.01, 0.005), 0.005, c(0.01, 0.005),
    0.005, 0.005, 0.001, 0.005),
  ZIs = rep(TRUE, 8),
  savegbms = rep(FALSE, 8),
  varints = rep(FALSE, 8),
  RSBs = rep(FALSE, 8),
  BnWs = rep(FALSE, 8),
  zeroes = rep(FALSE, 8))
```

gbm.factorplot	<i>Creates ggplots of marginal effect for factorial variables from plot.gbm in gbm.auto.</i>
----------------	--

Description

Creates an additional plot to those created by gbm.plot within gbm.auto. Can also take Bin/Gaus_Best_line.csv or similar csvs directly. Allows changing of x axis levels and all ggplot and ggsave params.

Usage

```
gbm.factorplot(
  x,
  factorplotlevels = NULL,
  ggplot2guideaxisangle = 0,
  ggplot2labsx = "",
  ggplot2labsy = "Marginal Effect",
  ggplot2axistext = 1.5,
  ggplot2axistitle = 2,
  ggplot2legendtext = 1,
  ggplot2legendtitle = 1.5,
  ggplot2legendtitlealign = 0,
  ggplot2plotbackgroundfill = "white",
  ggplot2plotbackgroundcolour = "grey50",
  ggplot2striptextx = 2,
```



```

ggplot2panelbordercolour = "black",
ggplot2panelborderfill = NA,
ggplot2panelborderlinewidth = 1,
ggplot2legendspacingx = grid::unit(0, "cm"),
ggplot2legendbackground = ggplot2::element_blank(),
ggplot2panelbackgroundfill = "white",
ggplot2panelbackgroundcolour = "grey50",
ggplot2panelgridcolour = "grey90",
ggplot2legendkey = ggplot2::element_blank(),
ggsavefilename = paste0(lubridate::today(), "_Categorical-variable.png"),
ggsaveplot = ggplot2::last_plot(),
ggsavedevice = "png",
ggsavepath = "",
ggsavescale = 2,
ggsavewidth = 10,
ggsaveheight = 4,
ggsaveunits = "in",
ggsavedpi = 300,
ggsavelimitsize = TRUE,
...
)

```

Arguments

x	Input data.frame or tibble or csv (full file address including .csv) to read, must be a categorical variable.
factorplotlevels	Character vector of the variable's levels to reorder the x axis by, all must match those in the first column of the csv exactly. Default NULL orders from high to low Y value.
ggplot2guideaxisangle	Default 0. Set at e.g. 90 to rotate.
ggplot2labsx	Default: "".
ggplot2labsy	Default: "Marginal Effect".
ggplot2axistext	Default: 1.5.
ggplot2axistitle	Default: 2.
ggplot2legendtext	Default: 1.
ggplot2legendtitle	Default: 1.5.
ggplot2legendtitlealign	Default: 0, # otherwise effect type title centre aligned for some reason.
ggplot2plotbackgroundfill	Default: "white", white background.
ggplot2plotbackgroundcolour	Default: "grey50", background lines.

```

ggplot2striptextx
    Default: 2.
ggplot2panelbordercolour
    Default: "black".
ggplot2panelborderfill
    Default: NA.
ggplot2panelborderlinewidth
    Default: 1.
ggplot2legendspacingx
    Default: unit(0, "cm"), # compress spacing between legend items, this is min.
ggplot2legendbackground
    Default: ggplot2::element_blank().
ggplot2panelbackgroundfill
    Default: "white".
ggplot2panelbackgroundcolour
    Default: "grey50".
ggplot2panelgridcolour
    Default: "grey90".
ggplot2legendkey
    Default: ggplot2::element_blank().
ggsavefilename Default: paste0(saveloc, lubridate::today(), "_SankeyAlluvial_EMT.SoEv-EfTyp_Col-
EfSz.png").
ggsaveplot      Default: last_plot().
ggsavedevice    Default: "png".
ggsavepath      Default: "".
ggsavescale     Default: 2.
ggsavewidth     Default: 10.
ggsaveheight    Default: 4.
ggsaveunits     Default: "in".
ggsavedpi       Default: 300.
ggsavelimitsize
    Default: TRUE.
...            Allow params to be called from higher function esp gbm.auto.

```

Details

```
‘r lifecycle::badge("experimental")
```

Value

Factorial ggplot saved with users preferred location and name.

Author(s)

Simon Dedman, <simondedman@gmail.com>

gbm.lmplots

*Plot linear models for all expvar against the resvar***Description**

Loops the lmpplot function, shows linear model plots for all expvar against the resvar. Good practice to do this before running gbm.auto so you have a sense of the basic relationship of the variables.

Usage

```
gbm.lmplots(
  samples = NULL,
  expvar = NULL,
  resvar = NULL,
  expvarnames = NULL,
  resvarname = NULL,
  savedir = NULL,
  plotname = NULL,
  pngtype = c("cairo-png", "quartz", "Xlib"),
  r2line = TRUE,
  pointtext = FALSE,
  pointlabs = resvar,
  pointcol = "black",
  ...
)
```

Arguments

<code>samples</code>	Explanatory and response variables to predict from. Keep col names short (~17 characters max), no odd characters, spaces, starting numerals or terminal periods. Spaces may be converted to periods in directory names, underscores won't. Can be a subset of a large dataset.
<code>expvar</code>	Vector of names or column numbers of explanatory variables in 'samples': c(1,3,6) or c("Temp","Sal"). No default.
<code>resvar</code>	Name or column number(s) of response variable in samples: 12, c(1,4), "Rockfish". No default. Column name is ideally species name.
<code>expvarnames</code>	Vector of names same length as expvar, if you want nicer names.
<code>resvarname</code>	Single character object, if you want a nicer resvar name.
<code>savedir</code>	Save location, end with "/".
<code>plotname</code>	Character vector of plot names else expvarnames else expvar will be used.
<code>pngtype</code>	Filetype for png files, alternatively try "quartz" on Mac.
<code>r2line</code>	Plot rsquared trendline, default TRUE.
<code>pointtext</code>	Label each point? Default FALSE.
<code>pointlabs</code>	Point labels, defaults to resvar value.
<code>pointcol</code>	Points colour, default "black".
<code>...</code>	Allows controlling of text label params e.g. adj cex &.

Details

Errors and their origins:

Value

Invisibly saves png plots into savedir.

Author(s)

Simon Dedman, <simondedman@gmail.com>

gbm.loop

Calculate Coefficient Of Variation surfaces for gbm.auto predictions

Description

Bagging introduces stochasticity which can result in sizeable variance in output predictions by gbm.auto for small datasets. This function runs a user- specified number of loops through the same gbm.auto parameter combinations and calculates the Coefficient Of Variation in the predicted abundance scores for each site aka cell. This can be mapped, to spatially demonstrate the output variance range.

Usage

```
gbm.loop(
  loops = 10,
  savedir = tempdir(),
  savecsv = TRUE,
  calcpreds = TRUE,
  varmap = TRUE,
  measure = "CPUE",
  cleanup = FALSE,
  grids = NULL,
  samples,
  expvar,
  resvar,
  randomvar = FALSE,
  tc = c(2),
  lr = c(0.01),
  bf = 0.5,
  n.trees = 50,
  ZI = "CHECK",
  fam1 = c("bernoulli", "binomial", "poisson", "laplace", "gaussian"),
  fam2 = c("gaussian", "bernoulli", "binomial", "poisson", "laplace"),
  simp = TRUE,
  gridslat = 2,
  gridslon = 1,
```

```

    multiplot = FALSE,
    cols = grey.colors(1, 1, 1),
    linesfiles = TRUE,
    smooth = FALSE,
    savegbm = FALSE,
    loadgbm = NULL,
    varint = FALSE,
    map = TRUE,
    shape = NULL,
    RSB = FALSE,
    BnW = FALSE,
    alerts = FALSE,
    pngtype = c("cairo-png", "quartz", "Xlib"),
    gaus = TRUE,
    MLEvaluate = TRUE,
    runautos = TRUE,
    Min.Inf = NULL,
    ...
)

```

Arguments

loops	The number of loops required, integer.
savendir	Save outputs to a temporary directory (default) else change to current directory e.g. "/home/me/folder". Do not use getwd() here.
savecsv	Save coefficients of variation in simple & extended format.
calcpreds	Calculate coefficients of variation of predicted abundance?
varmap	Create a map of the coefficients of variation outputs?
measure	Map legend, coefficients of variation of what? Default CPUE.
cleanup	Remove gbm.auto-generated directory each loop? Default FALSE.
grids	See gbm.auto help for all subsequent params.
samples	See gbm.auto help.
expvar	See gbm.auto help.
resvar	See gbm.auto help.
randomvar	See gbm.auto help.
tc	See gbm.auto help.
lr	See gbm.auto help.
bf	See gbm.auto help.
n.trees	See gbm.auto help.
ZI	See gbm.auto help. Choose one.
fam1	See gbm.auto help. Choose one.
fam2	See gbm.auto help. Choose one.
simp	See gbm.auto help.

<code>gridslat</code>	See <code>gbm.auto</code> help.
<code>gridslon</code>	See <code>gbm.auto</code> help.
<code>multiplot</code>	See <code>gbm.auto</code> help. Default False
<code>cols</code>	See <code>gbm.auto</code> help.
<code>linesfiles</code>	See <code>gbm.auto</code> help; TRUE or linesfiles calculations fail.
<code>smooth</code>	See <code>gbm.auto</code> help.
<code>savegbm</code>	See <code>gbm.auto</code> help.
<code>loadgbm</code>	See <code>gbm.auto</code> help.
<code>varint</code>	See <code>gbm.auto</code> help.
<code>map</code>	See <code>gbm.auto</code> help.
<code>shape</code>	See <code>gbm.auto</code> help.
<code>RSB</code>	See <code>gbm.auto</code> help.
<code>BnW</code>	See <code>gbm.auto</code> help.
<code>alerts</code>	See <code>gbm.auto</code> help; default FALSE as frequent use can crash RStudio.
<code>pngtype</code>	See <code>gbm.auto</code> help. Choose one.
<code>gaus</code>	See <code>gbm.auto</code> help.
<code>MLEvaluate</code>	See <code>gbm.auto</code> help.
<code>runautos</code>	Run <code>gbm.autos</code> , default TRUE, turn off to only collate numbered-folder results.
<code>Min.Inf</code>	Dummy param for package testing for CRAN, ignore.
<code>...</code>	Additional params for <code>gbm.auto</code> sub-functions including <code>gbm.step</code> .

Details

Thanks to a 2023 improvement to `gbm.auto` and `gbm.loop`,

Value

Returns a data frame of lat, long, 1 predicted abundance per loop, and a final variance score per cell.

Author(s)

Simon Dedman, <simondedman@gmail.com>

Examples

```
# Not run: downloads and saves external data.
library("gbm.auto")
data(grids) # load grids
data(samples) # load samples
gbmloopexample <- gbm.loop(loops = 2, samples = samples,
  grids = grids, expvar = c(4:10), resvar = 11, simp = F)
```

gbm.map

*Maps of predicted abundance from Boosted Regression Tree modelling***Description**

Generates maps from the outputs of gbm.step then Gbm.predict.grids, handled automatically within gbm.auto but can be run alone, and generates representativeness surfaces from the output of gbm.rsb.

Usage

```
gbm.map(
  x,
  y,
  z,
  byx = NULL,
  byy = NULL,
  grdfun = mean,
  mapmain = "Predicted CPUE (numbers per hour): ",
  species = "Response Variable",
  heatcolours = c("white", "yellow", "orange", "red", "brown4"),
  colournumber = 8,
  shape = NULL,
  landcol = "grey80",
  mapback = "lightblue",
  legendloc = "bottomright",
  legendtitle = "CPUE",
  lejback = "white",
  zero = TRUE,
  quantile = 1,
  byxout = FALSE,
  breaks = NULL,
  byxport = NULL,
  ...
)
```

Arguments

x	Vector of longitudes, from make.grid in mapplots; x. Order by this (descending) SECOND.
y	Vector of latitudes, from make.grid in mapplots; grids[,gridslat]. Order by this (descending) first.
z	Vector of abundances generated by gbm.predict.grids, from make.grid in mapplots; grids[,predabund].
byx	Longitudinal width of grid cell, from make.grid in mapplots. Autogenerated if left blank.

byy	Latitudinal height of grid cell, from make.grid in mapplots. Autogenerated if left blank.
grdfun	make.grid operand for ≥ 2 values per cell. Default:mean, other options: sum prod min max sd se var.
mapmain	Plot title, has species value appended. Default "Predicted CPUE (numbers per hour): ".
species	Response variable name, from basemap in mapplots; names(samples[i]). Defaults to "Response Variable".
heatcolours	Vector for abundance colour scale, defaults to the heatcol from legend.grid and draw.grid in mapplots which is c("white", "yellow", "orange", "red", "brown4").
colournumber	Number of colours to spread heatcol over, default:8.
shape	Basemap shape to draw, from draw.shape in mapplots. Defaults to NULL which calls gbm.basemap to generate it for you. First read in a shp file e.g. myshape <- sf::st_read(dsn = paste0(savename, ".shp"), layer = savename, quiet = TRUE), then use shape = myshape.
landcol	Colour for 'null' area of map (for marine plots, this is land), from draw.shape in mapplots. Default "grey80" (light grey).
mapback	Basemap background colour, defaults to lightblue (ocean for marine plots).
legendloc	Location on map of legend box, from legend.grid in mapplots, default bottom-right.
legendtitle	The metric of abundance, e.g. CPUE for fisheries, from legend.grid in mapplots. Default "CPUE".
lejback	Background colour of legend, from legend.grid in mapplots. Default "white".
zero	Force include 0-only bin in breaks.grid and thus legend? Default TRUE.
quantile	Set max quantile of data to include in bins, from breaks.grid in mapplots; lower to e.g. 0.975 cutoff outliers; default 1.
byxout	Export byx to use elsewhere? Default:FALSE.
breaks	Vector of breakpoints for colour scales; default blank, generated automatically.
byxport	Dummy param for package testing for CRAN, ignore.
...	Additional arguments for legend.grid's ... which passes to legend.

Details

[Superseded] Superseded by gbm.mapsf on 2023-08-07, but still works.

Errors and their origins:

Error in seq.default(xlim[1], xlim[2], by = byx):wrong sign in 'by' argument Check that your lat & long columns are the right way around. Ensure grids data are gridded, i.e. they are in a regular pattern of same/similar lines of lat/lon, even if they're missing sections.

Suggested parameter values: z = rsbdf[, "Unrepresentativeness"]

mapmain = "Unrepresentativeness: "

legendtitle = "UnRep 0-1"

Value

Species abundance maps using data provided by gbm.auto, and Representativeness Surface Builder maps using data provided by gbm.rsb, to be run in a png/par/gbm.map/dev.off sequence.

Author(s)

Simon Dedman, <simondedman@gmail.com>

Hans Gerritsen

Examples

```
# Not run: downloads and saves external data.
# Suggested code for outputting to png:
data(grids)
# set working directory somewhere suitable
png(filename = "gbmmap.png", width = 7680, height = 7680, units = "px",
pointsize = 192, bg = "white", res = NA, family = "", type = "cairo-png")
par(mar = c(3.2,3,1.3,0), las = 1, mgp = c(2.1,0.5,0), xpd = FALSE)
gbm.map(x = grids[, "Longitude"], y = grids[, "Latitude"], z = grids[, "Effort"]
, species = "Effort")
dev.off()
```

gbm.mapsf

Maps of predicted abundance from Boosted Regression Tree modelling

Description

Generates maps from the outputs of gbm.step then Gbm.predict.grids, handled automatically within gbm.auto but can be run alone, and generates representativeness surfaces from the output of gbm.rsb.

Usage

```
gbm.mapsf(
  predabund = NULL,
  predabundlon = 2,
  predabundlat = 1,
  predabundpreds = 3,
  myLocation = NULL,
  trim = TRUE,
  trimfivepct = FALSE,
  scale100 = FALSE,
  gmapsAPI = NULL,
  mapsource = "google",
  googlemap = TRUE,
  maptype = "satellite",
```

```

darkenproportion = 0,
mapzoom = NULL,
shape = NULL,
expandfactor = 0,
colourscale = "viridis",
colorscale = NULL,
heatcolours = c("white", "yellow", "orange", "red", "brown4"),
colournumber = 8,
colourscalelimits = NULL,
colourscalebreaks = NULL,
colourscalelabels = NULL,
colourscaleexpand = NULL,
studyspecies = "MySpecies",
plottitle = paste0("Predicted abundance of ", studyspecies),
plotsubtitle = "CPUE",
legendtitle = "CPUE",
plotcaption = paste0("gbm.auto::gbm.mapsf, ", lubridate::today()),
axisxlabel = "Longitude",
axisylabel = "Latitude",
legendposition = c(0.05, 0.15),
fontsize = 12,
fontfamily = "Times New Roman",
filesavename = paste0(lubridate::today(), "_", studyspecies, "_", legendtitle, ".png"),
savedir = tempdir(),
receiverlats = NULL,
receiverlons = NULL,
receivernames = NULL,
receiverrange = NULL,
recpointscol = "black",
recpointsfill = "white",
recpointsalpha = 0.5,
recpointssize = 1,
recpointsshape = 21,
recbufcol = "grey75",
recbuffill = "grey",
recbufalpha = 0.5,
reclabcol = "black",
reclabfill = NA,
reclabnudgex = 0,
reclabnudgexy = -200,
reclabpad = 0,
reclabrad = 0.15,
reclabbord = 0
)

```

Arguments

predabund	Predicted abundance data frame produced by gbm.auto (Abundance_Preds_only.csv), with Latitude, Longitude, and Predicted Abundance columns. Default NULL.
-----------	--

	You need to read the csv in R if not already present as an object in the environment.
predabundlon	Longitude column number. Default 2.
predabundlat	Latitude column number. Default 1.
predabundpreds	Predicted abundance column number, default 3.
myLocation	Location for extents, format c(xmin, ymin, xmax, ymax). Default NULL, extents autocreated from data.
trim	Remove NA & <=0 values and crop to remaining data extents? Default TRUE.
trimfivepct	Replace anything < 5% of the max value (i.e. < 95% UD contour in home range analysis) with NA since it won't be drawn (for movegroup dBBMMs). Default FALSE.
scale100	Scale Predicted Abundance to 100? Default FALSE.
gmapsAPI	Enter your Google maps API here, quoted character string. Default NULL.
mapsource	Source for ggmap::get_map; uses Stamen as fallback if no Google Maps API present. Options: "google", "stamen", "gbm.basemap". Default "google". Using "gbm.basemap" requires one to have run that function already, and enter its location using the shape parameter below.
googlemap	If pulling basemap from Google maps, this sets expansion factors since Google Maps tiling zoom setup doesn't align to myLocation extents. Default TRUE.
maptype	Type of map for ggmap::get_map param maptype. Options: Google mapsource: "terrain", "terrain-background", "satellite", "roadmap", "hybrid". Stamen mapsource: "terrain", "terrain-background", "terrain-labels", "terrain-lines", "watercolor", "toner", "toner-2010", "toner-2011", "toner-background", "toner-hybrid", "toner-labels", "toner-lines", "toner-lite".
darkenproportion	Amount to darken the google/stamen basemap, 0-1. Default 0.
mapzoom	Highest number = zoomed in. Google: 3 (continent) - 21 (building). stamen: 0-18. Default 9.
shape	If mapsource is "gbm.basemap", enter the full path to gbm.basemaps downloaded map, typically Crop_Map.shp, including the .shp. Default NULL. Can also name an existing object in the environment, read in with sf::st_read.
expandfactor	Extents expansion factor for basemap. default 0.
colourscale	Scale fill colour scheme to use, default "viridis", other option is "gradient".
colorscale	Scale fill colour scheme to use, default NULL, populating this will overwrite colourscale.
heatcolours	Vector of colours if gradient selected for colourscale, defaults to heatmap theme.
colournumber	Number of colours to spread heatcolours over, if gradient selected for colourscale. Default 8.
colourscalelimits	Colour scale limits, default NULL, vector of 2, e.g. c(0, 0).
colourscalebreaks	Colour scale breaks, default NULL.

colourscalelabels	Colour scale labels, default NULL, must match number of breaks.
colourscaleexpand	Colour scale expand, default NULL, vector of 2, e.g. c(0, 0).
studyspecies	Name of your study species, appears in plot title and savename. Default "MySpecies".
plottitle	Title of the resultant plot, default paste0("Predicted abundance of ", studyspecies).
plotsubtitle	Plot subtitle, default ""CPUE". Can add the n of your individuals.
legendtitle	Legend title, default "CPUE".
plotcaption	Plot caption, default "gbm.auto::gbm.mapsf" + today's date.
axisxlabel	Default "Longitude".
axisylabel	Default "Latitude".
legendposition	Vector of 2, format c(1,2), Proportional distance of (middle?) of legend box from L to R, percent distance from Bottom to Top. Values 0 to 1. Default c(0.05, 0.15).
fontsize	Font size, default 12.
fontfamily	= Font family, default "Times New Roman".
filesavename	File savename, default today's date + studyspecies + legendtitle.
savendir	Save outputs to a temporary directory (default) else change to current directory e.g. "/home/me/folder". Do not use getwd() here. No terminal slash. E.g. paste0(movegroupsavendir, "Plot/") .
receiverlats	Vector of latitudes for receivers to be plotted.
receiverlons	Vector of longitudes for receivers to be plotted. Same length as receiverlats.
receivernames	Vector of names for receivers to be plotted. Same length as receiverlats.
receiverrange	Single (will be recycled), or vector (same length as receiverlats) of detection ranges in metres for receivers to be plotted. If you have a max and a (e.g.) 90 percent detection range, probably use max.
recpointscol	Colour of receiver centrepoint outlines. Default "black".
recpointsfill	Colour of receiver centrepoint fills. Default "white".
recpointsalph	Alpha value of receiver centrepoint fills, 0 (invisible) to 1 (fully visible). Default 0.5.
recpointssize	Size of receiver points. Default 1.
recpointsshape	Shape of receiver points, default 21, circle with outline and fill.
recbufcol	Colour of the receiver buffer circle outlines. Default "grey75"
recbuffill	Colour of the receiver buffer circle fills. Default "grey".
recbufalpha	Alpha value of receiver buffer fills, 0 (invisible) to 1 (fully visible). Default 0.5.
reclabcol	Receiver label text colour. Default "black".
reclabfill	Receiver label fill colour, NA for no fill. Default NA.
reclabnudgeX	Receiver label offset nudge in X dimension. Default 0.
reclabnudgeY	Receiver label offset nudge in Y dimension. Default -200.
reclabpad	Receiver label padding in lines. Default 0.
reclabrad	Receiver label radius in lines. Default 0.15.
reclabbord	Receiver label border in mm. Default 0.

Details

Error in seq.default(xlim[1], xlim[2], by = byx):wrong sign in 'by' argument Check that your lat & long columns are the right way around. Ensure grids (predabund) data are gridded, i.e. they are in a regular pattern of same/similar lines of lat/lon, even if they're missing sections.

Suggested parameter values: z = rsbdf[, "Unrepresentativeness"]

mapmain = "Unrepresentativeness: "

legendtitle = "UnRep 0-1"

How to get Google map basemaps:

(from <https://www.youtube.com/watch?v=O5cUoVpVUjU>):

1. Sign up with dev console: a. You must enter credit card details, but won't be charged if your daily API requests stay under the limit. b. Follow the link: <https://console.cloud.google.com/projectselector2/apis/dashboard>. c. Sign up for Google cloud account (it may auto populate your current gmail), click agree and continue. d. Click the navigation email in the top left corner and click on Billing. e. Create a billing account – they will NOT auto charge after trial ends. f. Enter information, click on 'start my free trial'. They may offer a free credit for trying out their service. More pricing details: <https://mapsplatform.google.com/pricing/>. g. Click "Select a Project" then "New project" in the top right corner. h. Enter Project Name, leave Location as is, click "Create". i. You should now see your project name at the top, where the drop-down menu is.
2. Enable Maps and Places API: a. Click 'Library' on the left. b. In the search field type "Maps". c. Scroll down, click "Maps Java Script API". d. Click Enable. e. Click 'Library' again, search "Places", click on "Places API". f. Click Enable.
3. Create Credentials for API Key: a. Return to 'APIs & Services' page. b. Click on Credentials. c. At the top click 'Create Credentials > API Key'. d. API key should pop up with option to copy it. e. You can restrict the key if you want by following steps 4 & 5 here: <https://www.youtube.com/watch?v=O5cUoVpVUjU&t=232s>

Value

Species abundance maps using data provided by gbm.auto, and Representativeness Surface Builder maps using data provided by gbm.rsb, to be run in a png/par/gbm.map/dev.off sequence.

Author(s)

Simon Dedman, <simondedman@gmail.com>

Examples

```
# Not run
```

gbm.rsb

*Representativeness Surface Builder***Description**

Loops through explanatory variables comparing their histogram in 'samples' to their histogram in 'grids' to see how well the explanatory variable range in samples represents the range being predicted to in grids. Assigns a representativeness score per variable per site in grids, and takes the average score per site if there's more than 1 expvar. Saves this to a CSV; it's plotted by gbm.map if called in gbm.auto. This shows you which areas have the most and least representative coverage by samples, therefore where you can have the most/least confidence in the predictions from gbm.predict.grids. Can be called directly, and choosing a subset of expvars allows one to see their individual / collective representativeness.

Usage

```
gbm.rsb(samples, grids, expvarnames, gridslat, gridslon)
```

Arguments

<code>samples</code>	Data frame with response and explanatory variables.
<code>grids</code>	Data frame of (more/different) explanatory variables and no response variable, to be predicted to by gbm.predict.grids.
<code>expvarnames</code>	Vector of column names of explanatory variables being tested. Can be length 1. Names must match in samples and grids.
<code>gridslat</code>	Column number for latitude in 'grids'.
<code>gridslon</code>	Column number for longitude in 'grids'.

Value

Gridded data table of representativeness values which is then mapped with gbm.map and also saved as a csv

Author(s)

Simon Dedman, <simondedman@gmail.com>

Examples

```
data(samples)
data(grids)
rsbdf_bin <- gbm.rsb(samples, grids, expvarnames = names(samples[c(4:8, 10)])
, gridslat = 2, gridslon = 1)
```

gbm.step.sd	<i>Function to assess optimal no of boosting trees using k-fold cross validation</i>
-------------	--

Description

SD fork of dismo's gbm.step to add evaluation metrics like d.squared and rmse. J. Leathwick and J. Elith - 19th September 2005, version 2.9. Function to assess optimal no of boosting trees using k-fold cross validation. Implements the cross-validation procedure described on page 215 of Hastie T, Tibshirani R, Friedman JH (2001) The Elements of Statistical Learning: Data Mining, Inference, and Prediction Springer-Verlag, New York.

Usage

```
gbm.step.sd(
  data,
  gbm.x,
  gbm.y,
  offset = NULL,
  fold.vector = NULL,
  tree.complexity = 1,
  learning.rate = 0.01,
  bag.fraction = 0.75,
  site.weights = rep(1, nrow(data)),
  var.monotone = rep(0, length(gbm.x)),
  n.folds = 10,
  prev.stratify = TRUE,
  family = "bernoulli",
  n.trees = 50,
  step.size = n.trees,
  max.trees = 10000,
  tolerance.method = "auto",
  tolerance = 0.001,
  plot.main = TRUE,
  plot.folds = FALSE,
  verbose = TRUE,
  silent = FALSE,
  keep.fold.models = FALSE,
  keep.fold.vector = FALSE,
  keep.fold.fit = FALSE,
  ...
)
```

Arguments

data	The input dataframe.
gbm.x	The predictors.

gbm.y	The response.
offset	Allows an offset to be specified.
fold.vector	Allows a fold vector to be read in for CV with offsets,.
tree.complexity	Sets the complexity of individual trees.
learning.rate	Sets the weight applied to individual trees.
bag.fraction	Sets the proportion of observations used in selecting variables.
site.weights	Allows varying weighting for sites.
var.monotone	Restricts responses to individual predictors to monotone.
n.folds	Number of folds.
prev.stratify	Prevalence stratify the folds - only for p/a data.
family	Family - bernoulli (=binomial), poisson, laplace or gaussian.
n.trees	Number of initial trees to fit.
step.size	Numbers of trees to add at each cycle.
max.trees	Max number of trees to fit before stopping.
tolerance.method	Method to use in deciding to stop - "fixed" or "auto".
tolerance	Tolerance value to use - if method == fixed is absolute, if auto is multiplier * total mean deviance.
plot.main	Plot hold-out deviance curve.
plot.folds	Plot the individual folds as well.
verbose	Control amount of screen reporting.
silent	To allow running with no output for simplifying model).
keep.fold.models	Keep the fold models from cross validation.
keep.fold.vector	Allows the vector defining fold membership to be kept.
keep.fold.fit	Allows the predicted values for observations from CV to be kept.
...	Allows for any additional plotting parameters.

Details

Divides the data into 10 subsets, with stratification by prevalence if required for pa data then fits a gbm model of increasing complexity along the sequence from n.trees to n.trees + (n.steps * step.size) calculating the residual deviance at each step along the way after each fold processed, calculates the average holdout residual deviance and its standard error then identifies the optimal number of trees as that at which the holdout deviance is minimised and fits a model with this number of trees, returning it as a gbm model along with additional information from the cv selection process.

D squared is $1 - (cv.dev / total.deviance)$. Abeare thesis: For each of the fitted models, the pseudo-R², or D², or Explained Deviance, was calculated for comparison, where: $D^2 = 1 - (residual\ deviance / total\ deviance)$.

requires gbm library from Cran requires roc and calibration scripts of J Elith requires calc.deviance script of J Elith/J Leathwick

Value

GBM models using gbm as the engine.

gbm.subset	<i>Subset gbm.auto input datasets to 2 groups using the partial deviance plots</i>
------------	--

Description

Set your working directory to the output folder of a gbm.auto/gbm.loop run. This function returns the variable value corresponding to the 0 value on the lineplots, which should be the optimal place to split the dataset into 2 subsets, low and high, IF the relationship doesn't cross 0 more than once. Function is similarly useful to quickly get the 0-point value in these cases, i.e. where values below are detrimental, values above beneficial (check plots though)

Usage

```
gbm.subset(x, fams = c("Bin", "Gaus"), loop = FALSE)
```

Arguments

x	Vector of variable names.
fams	Vector of statistical data distribution family names to be modelled by gbm.
loop	Is the folder a gbm.loop output?

Details

loop varnames are BinLineLoop_VAR.csv & GausLineLoop_VAR.csv normal varnames are Bin_Best_line_VAR.csv & Gaus_Best_line_VAR.csv

Just use average between the last negative & first positive point unless any points fall on zero

Value

a list of breakpoint values which datasets can be subsetted using.

Author(s)

Simon Dedman, <simondedman@gmail.com>

Examples

```
# Not run: requires completed gbm.auto run.
# having run gbm.auto (with linesfiles=TRUE), set working directory there
data(samples)
gbm.subset(x = names(samples[c(4:8, 10)]), fams = c("Bin", "Gaus"))
```

gbm.valuemap

Decision Support Tool that generates (Marine) Protected Area options using species predicted abundance maps

Description

Scales response variable data, maps a user-defined explanatory variable to be avoided, e.g. fishing effort, combines them into a map showing areas to preferentially close. Bpa, the precautionary biomass required to protect the spawning stock, is used to calculate MPA size. MPA is then grown to add subsequent species starting from the most conservationally at-risk species, resulting in one MPA map per species, and a multicolour MPA map of all. All maps list the percentage of the avoid-variables total that is overlapped by the MPA in the map legend.

Usage

```
gbm.valuemap(
  dbase,
  loncolno = 1,
  latcolno = 2,
  goodcols,
  badcols,
  conservecol = NULL,
  plotthis = c("good", "bad", "both", "close"),
  maploops = c("Combo", "Biomass", "Effort", "Conservation"),
  savedir = tempdir(),
  savethis = TRUE,
  HRMSY = 0.15,
  goodweight = NULL,
  badweight = NULL,
  m = 1,
  alerts = TRUE,
  BnW = TRUE,
  shape = NULL,
  pngtype = c("cairo-png", "quartz", "Xlib"),
  byxport = NULL,
  ...
)
```

Arguments

dbase	Data.frame to load. Expects Lon, Lat & data columns: predicted abundances, fishing effort etc. E.g.: Abundance_Preds_All.csv from gbm.auto.
loncolno	Column number in dbase which has longitudes.
latcolno	Column number in dbase which has latitudes.
goodcols	Which column numbers are abundances (where higher = better)? List them in order of highest conservation importance first e.g. c(3,1,2,4). Either numeric column number or quoted character column name.

badcols	Which col no.s are 'negative' e.g. fishing (where higher = worse)? Either numeric column number or quoted character column name.
conservecol	Conservation column, from gbm.cons.
plotthis	Vector of variable types to plot. Delete any, or all w/ NULL.
maploops	Vector of sort loops to run. See Dedman et al 2017 "Towards a flexible Decision Support Tool for MSY-based Marine Protected Area design for skates and rays"; https://academic.oup.com/icesjms/article/74/2/576/2669563 . All 4 options create a total MPA which conserves Bpa, but in different ways: Biomass closes areas of high biomass first. Effort closes areas of high fisheries area last. Combo strikes a balance between the two, and you can change the default 1:1 balance with goodweight and badweight parameters. Conservation uses the output of gbm.cons to prioritise closure of areas of high conservation value, which may not be identical to areas of highest biomass.
savedir	Save outputs to a temporary directory (default) else change to current directory e.g. "/home/me/folder". Do not use getwd() here.
savethis	Export all data as csv?
HRMSY	Maximum percent of each goodcols stock which can be removed yearly, as decimal (0.15 = 15 pct). Must protect remainder: 1-HRMSY. Single number or vector. If vector, same order as goodcols. Required.
goodweight	Single/vector weighting multiple(s) for goodcols array.
badweight	Ditto for badcols array.
m	Multiplication factor for Bpa units. 1000 to convert tonnes to kilos, 0.001 kilos to tonnes. Assumedly the same for all goodcols.
alerts	Play sounds to mark progress steps.
BnW	Also produce greyscale images for print publications.
shape	Set coastline shapefile, else uses British Isles. Generate your own with gbm.basemap.
pngtype	File-type for png files, alternatively try "quartz" on Mac. Choose one.
byxport	Dummy param for package testing for CRAN, ignore.
...	Optional terms for gbm.map.

Details

Bpa is the volume of biomass under the 2D abundance surface e.g. predabund from gbm.auto. B (biomass), \times HRMSY (Fmsy proportion) = Bpa. You may be able to get Fmsy from stock assessments etc. maploops: explain concept of biomass vs effort, combo in the middle (default weighting 1:1 can change with good/badweight), and Conservation from gbm.cons.

Value

Species abundance, abundance vs avoid variable, and MPA maps per species and sort type, in b&w if set. CSVs of all maps if set.

Author(s)

Simon Dedman, <simondedman@gmail.com>

grids

*Data: Explanatory variables for rays in the Irish Sea***Description**

A dataset containing explanatory variables for environment, fishery and predators of rays including juveniles in the Irish Sea.

Usage

```
data(grids)
```

Format

A data frame with 378570 rows and 43 variables:

Longitude Decimal longitudes in the Irish Sea

Latitude Decimal latitudes in the Irish Sea

Depth Metres, decimal

Temperature Degrees, decimal

Salinity PPM

Current_Speed Metres per second at the seabed

Distance_to_Shore Metres, decimal

F_LPUE Commercial fishery LPUE in Kg/Hr

Scallop Average Kwh Scallop effort from logbooks, Marine Institute and MMO combined

MI_Av_E_Hr Average effort hours, Marine Institute Scallop VMS, 0.03 x 0.02 rectangles, all Irish Sea, 2006-14

MI_Av_LPUE Average scallop CPUE, Marine Institute Scallop VMS, 0.03 x 0.02 rectangles, all Irish Sea, 2006-14

MI_Sum_Liv Sum of live weight. Average scallop CPUE, Marine Institute Scallop VMS, 0.03 x 0.02 rectangles, all Irish Sea, 2006-14

Whelk MMO Whelk LPUE 2009-12, pivot, polygons to points

MmoAvScKwh MMO Scallop Effort 2009-12, pivot, polygons to points. ICES rectangles

HubDist map calc, distance of grid point to nearest datras point representing it (for preds)

Cod_C ICES IBTS CPUE of cod caught between 1994 - 2014 large enough to predate upon <= year 1 cuckoo rays

Cod_T As Cod_C for yr1 thornback rays

Cod_B As Cod_C for yr1 blonde rays

Cod_S As Cod_C for yr1 spotted rays

Haddock_C As Cod_C, haddock predating upon cuckoo rays

Haddock_T As Cod_C, haddock predating upon thornback rays

Haddock_B As Cod_C, haddock predating upon blonde rays
Haddock_S As Cod_C, haddock predating upon spotted rays
Plaice_C As Cod_C, plaice predating upon cuckoo rays
Plaice_T As Cod_C, plaice predating upon thornback rays
Plaice_B As Cod_C, plaice predating upon blonde rays
Plaice_S As Cod_C, plaice predating upon spotted rays
Whiting_C As Cod_C, whiting predating upon cuckoo rays
Whiting_T As Cod_C, whiting predating upon thornback rays
Whiting_B As Cod_C, whiting predating upon blonde rays
Whiting_S As Cod_C, whiting predating upon spotted rays
ComSkt_C As Cod_C, common skate predating upon cuckoo rays
ComSkt_T As Cod_C, common skate predating upon thornback rays
ComSkt_B As Cod_C, common skate predating upon blonde rays
ComSkt_S As Cod_C, common skate predating upon spotted rays
Blonde_C As Cod_C, blonde ray predating upon cuckoo rays
Blonde_T As Cod_C, blonde ray predating upon thornback rays
Blonde_S As Cod_C, blonde ray predating upon spotted rays
C_Preds All predator CPUEs combined for cuckoo rays
T_Preds All predator CPUEs combined for thornback rays
B_Preds All predator CPUEs combined for blonde rays
S_Preds All predator CPUEs combined for spotted rays
Effort Irish commercial beam trawler effort 2012

Author(s)

Simon Dedman, <simondedman@gmail.com>

Source

<http://oar.marine.ie/handle/10793/958>

 Juveniles

Data: Explanatory and response variables for 4 juvenile rays in the Irish Sea

Description

A dataset containing explanatory variables for environment, fishery and predators of juvenile rays in the Irish Sea, and the response variables, abundance CPUEs of cuckoo, thornback, blonde and spotted rays.

Usage

```
data(Juveniles)
```

Format

A data frame with 2136 rows and 46 variables:

Survey_StNo_HaulNo_Year Index column of combined Survey number, station number, haul number, and year

Latitude Decimal latitudes in the Irish Sea

Longitude Decimal longitudes in the Irish Sea

Depth Metres, decimal

Temperature Degrees, decimal

Salinity PPM

Current_Speed Metres per second at the seabed

Distance_to_Shore Metres, decimal

F_LPUE Commercial fishery LPUE in Kg/Hr

Scallop Average Kwh Scallop effort from logbooks, Marine Institute and MMO combined

MI_Av_E_Hr Average effort hours, Marine Institute Scallop VMS, 0.03 x 0.02 rectangles, all Irish Sea, 2006-14

MI_Av_LPUE Average scallop CPUE, Marine Institute Scallop VMS, 0.03 x 0.02 rectangles, all Irish Sea, 2006-14

MI_Sum_Liv Sum of live weight. Average scallop CPUE, Marine Institute Scallop VMS, 0.03 x 0.02 rectangles, all Irish Sea, 2006-14

Whelk MMO Whelk LPUE 2009-12, pivot, polygons to points

MmoAvScKwh MMO Scallop Effort 2009-12, pivot, polygons to points. ICES rectangles

Cod_C ICES IBTS CPUE of cod caught between 1994 - 2014 large enough to predate upon <= year 1 cuckoo rays

Cod_T As Cod_C for yr1 thornback rays

Cod_B As Cod_C for yr1 blonde rays

Cod_S As Cod_C for yr1 spotted rays

Haddock_C As Cod_C, haddock predating upon cuckoo rays
Haddock_T As Cod_C, haddock predating upon thornback rays
Haddock_B As Cod_C, haddock predating upon blonde rays
Haddock_S As Cod_C, haddock predating upon spotted rays
Plaice_C As Cod_C, plaice predating upon cuckoo rays
Plaice_T As Cod_C, plaice predating upon thornback rays
Plaice_B As Cod_C, plaice predating upon blonde rays
Plaice_S As Cod_C, plaice predating upon spotted rays
Whiting_C As Cod_C, whiting predating upon cuckoo rays
Whiting_T As Cod_C, whiting predating upon thornback rays
Whiting_B As Cod_C, whiting predating upon blonde rays
Whiting_S As Cod_C, whiting predating upon spotted rays
ComSkt_C As Cod_C, common skate predating upon cuckoo rays
ComSkt_T As Cod_C, common skate predating upon thornback rays
ComSkt_B As Cod_C, common skate predating upon blonde rays
ComSkt_S As Cod_C, common skate predating upon spotted rays
Blonde_C As Cod_C, blonde ray predating upon cuckoo rays
Blonde_T As Cod_C, blonde ray predating upon thornback rays
Blonde_S As Cod_C, blonde ray predating upon spotted rays
C_Preds All predator CPUEs combined for cuckoo rays
T_Preds All predator CPUEs combined for thornback rays
B_Preds All predator CPUEs combined for blonde rays
S_Preds All predator CPUEs combined for spotted rays
Cuckoo Numbers of juvenile cuckoo rays caught, standardised to 1 hour
Thornback Numbers of juvenile thornback rays caught, standardised to 1 hour
Blonde Numbers of juvenile blonde rays caught, standardised to 1 hour
Spotted Numbers of juvenile spotted rays caught, standardised to 1 hour

Author(s)

Simon Dedman, <simondedman@gmail.com>

implot

Plot linear model for two variables with R2 & P printed and saved

Description

Simple function to plot and name a linear model

Usage

```
implot(
  x,
  y,
  xname = "X variable",
  yname = "Y variable",
  pngtype = c("cairo-png", "quartz", "Xlib"),
  xlab = xname,
  ylab = yname,
  plotname = xname,
  r2line = TRUE,
  pointtext = FALSE,
  pointlabs = x,
  pointcol = "black",
  savedir = "",
  ...
)
```

Arguments

x	Explanatory variable data.
y	Response variable data.
xname	Variable name for plot header.
yname	Variable name for plot header.
pngtype	Filetype for png files, alternatively try "quartz" on Mac.
xlab	X axis label, parsed from xname unless specified.
ylab	Y axis label, parsed from yname unless specified.
plotname	Filename for png, parsed from xname unless specified.
r2line	Plot rsquared trendline, default TRUE.
pointtext	Label each point? Default FALSE.
pointlabs	Point labels, defaults to resvar value.
pointcol	Points colour, default "black".
savedir	Save location, end with "/".
...	Allows controlling of text label params e.g. adj cex &.

Details

Errors and their origins:

Value

Invisibly saves png plot into savedir.

Author(s)

Simon Dedman, <simondedman@gmail.com>

roc

roc

Description

Internal use only. Adapted from Ferrier, Pearce and Watson's code, by J.Elith , see: Hanley, J.A. & McNeil, B.J. (1982) The meaning and use of the area under a Receiver Operating Characteristic (ROC) curve. Radiology, 143, 29-36. Also Pearce, J. & Ferrier, S. (2000) Evaluating the predictive performance of habitat models developed using logistic regression. Ecological Modelling, 133, 225-245. This is the non-parametric calculation for area under the ROC curve, using the fact that a MannWhitney U statistic is closely related to the area. In dismo, this is used in the gbm routines, but not elsewhere (see evaluate).

Usage

```
roc(obsdat, preddat)
```

Arguments

obsdat	Observed data.
preddat	Predicted data.

Value

roc & calibration stats internally within gbm runs e.g. in gbm.auto.

Author(s)

Simon Dedman, <simondedman@gmail.com>

Examples

```
roc(obsdat = rbinom(100,size = 1, prob = 0.5), preddat = runif(100))
```

samples

Data: Numbers of 4 ray species caught in 2137 Irish Sea trawls, 1994 to 2014

Description

2244 capture events of cuckoo, thornback, spotted and blonde rays in the Irish Sea from 1994 to 2014 by the ICES IBTS, including explanatory variables: Length Per Unit Effort in that area by the commercial fishery, fishing effort by same, depth, temperature, distance to shore, and current speed at the bottom.

Usage

`data(samples)`

Format

A data frame with 2244 rows and 14 variables:

Survey_StNo_HaulNo_Year Index column of combined Survey number, station number, haul number, and year

Latitude Decimal latitudes in the Irish Sea

Longitude Decimal longitudes in the Irish Sea

Depth Metres, decimal

Temperature Degrees, decimal

Salinity PPM

Current_Speed Metres per second at the seabed

Distance_to_Shore Metres, decimal

F_LPUE Commercial fishery LPUE in Kg/Hr

Effort Irish commercial beam trawler effort 2012

Cuckoo Numbers of juvenile cuckoo rays caught, standardised to 1 hour

Thornback Numbers of juvenile thornback rays caught, standardised to 1 hour

Blonde Numbers of juvenile blonde rays caught, standardised to 1 hour

Spotted Numbers of juvenile spotted rays caught, standardised to 1 hour

Author(s)

Simon Dedman, <simondedman@gmail.com>

Source

<http://oar.marine.ie/handle/10793/958>

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