

Package ‘iSDM’

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Type Package

Title Invasive Species Distribution Modelling

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Description Functions for predicting and mapping potential and realized distributions of invasive species within the invaded range.

Depends R (>= 3.2.0)

Imports stats, utils, graphics, ade4, geometry, geosphere, maptools, MASS, pdist, raster, rgdal, sp, colorRamps, virtualspecies, rgl

License GPL-3

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NeedsCompilation no

Repository CRAN

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eSample

R function to perform environmental systematic sampling

Description

R function to perform environmental systematic sampling

Usage

```
eSample(envData, nExpect, plot=TRUE, saveShape=TRUE, nf, lowerLim, upperLim)
```

Arguments

envData	Either a SpatialPixelsDataFrame or a SpatialGridDataFrame as defined in package sp or a RasterStack or RasterBrick as defined in package raster. Note that this object can contain a mixture of variables type (quantitative, factor and ordered)
nExpect	Numeric number of desired sampling points. Note that sometime the exact expected number can not be reached because this function tries to find a grid that best matches the expected number
plot	A logic indicating whether or not you would like to have a graphical representation of the environmental systematic sampling (default=TRUE)
saveShape	A logic indicating whether or not you want to save the geographical coordinates of the pixels corresponding to this environmental systematic sampling as a shapefile format (default=TRUE). The generated shapefile will be saved in your working directory
nf	A numeric value indicating the number of ordination axes retained in the analysis, this function manages currently only 2 or 3 dimensions
lowerLim	Numeric value of probability in [0,1] that can be used to produce sample quantiles corresponding to the given lower limit probability. This may be used to reduce the effect of extremes observations
upperLim	Numeric value of probability in [0,1] that can be used to produce sample quantiles corresponding to the given upper limit probability. This may be used to reduce the effect of extremes observations

Details

The environmental systematic sampling consists in reducing the environmental space into 2 or 3 dimensions using an ordination method (the dudi.mixt method in ade4 package) as a first step. Thereafter convex hull will be created around the reduced environmental space. Then, a multidimensional grid will be created inside the convex hull. The obtained grid represents the perfect configuration required to adequately survey the environmental space. The last step is subsequently to seek the closest pixel to this ideal configuration. This is achieved by searching the nearest neighbour between each grid point and each pixel in the environmental space based on Euclidean distances

Value

Returns a list containing 4 objects: GeoSamples (The geographical coordinates of the pixels corresponding to the environmental systematic sampling); EnvSamples(The coordinates in the reduced environmental space of pixels corresponding to the environmental systematic sampling); EnvGrid (The coordinates of the grid in the reduced environmental space)

References

Tarek Hattab, Carol Ximena Garzon Lopez, Michael Ewald, Sandra Skowronek, Raf Aerts, Helene Horen, Boris Brasseur, Emilie Gallet-Moron, Fabien Spicher, Guillaume Decocq, Hannes Feilhauer, Olivier Honnay, Pieter Kempeneers, Sebastian Schmidlein, Ben Somers, Ruben Van De Kerchove, Duccio Rocchini and Jonathan Lenoir (2017). A unified framework to model the potential and realized distributions of invasive species within the invaded range. Diversity and Distributions. DOI: 10.1111/ddi.12566

See Also

<<http://diarsproject.github.io/DIARS/SamplingDesign.html>>

Examples

```
## Not run:
library(raster)
envData<-getData('worldclim', var='bio', res=10)
envData<-crop(envData,extent(-10,45,20,75))
plot(envData)
par(mfrow=c(1,2))
Mysampling1<-eSample(envData,nExpect=50,plot=TRUE,
saveShape=TRUE,nf=2,lowerLim=0.00001,upperLim=0.99999)
plot(envData[[1]])
plot(Mysampling1[[1]],add=TRUE,col=2,pch=19)
par(mfrow=c(1,2))
Mysampling2<-eSample(envData,nExpect=50,plot=TRUE,
saveShape=TRUE,nf=2,lowerLim=0.1,upperLim=0.9)
plot(envData[[1]])
plot(Mysampling2[[1]],add=TRUE,col=2,pch=19)
Mysampling3<-eSample(envData,nExpect=50,plot=TRUE,
saveShape=TRUE,nf=3,lowerLim=0.001,upperLim=0.999)
plot(envData[[1]])
plot(Mysampling3[[1]],add=TRUE,col=2)

## End(Not run)
```

iForce

R function to compute a negative exponential dispersal kernel

Description

R function to compute a negative exponential dispersal kernel

Usage

```
iForce(occData, coords, a, envData, binary=TRUE, longlat=NULL)
```

Arguments

occData	Either a <code>SpatialPointsDataFrame</code> as defined in package <code>sp</code> , a <code>data.frame</code> or a matrix object containing species data
coords	Optional 2 columns matrix containing the X and Y coordinates of <code>occData</code> (only consider if <code>occData</code> is a <code>data.frame</code> or a matrix object)
a	An integer between 0 and 1 that controls the form of the dispersal kernel
envData	Either a <code>SpatialPixelsDataFrame</code> or a <code>SpatialGridDataFrame</code> as defined in package <code>sp</code> or a <code>Raster</code> or <code>RasterLayer</code> as defined in package <code>raster</code> . This object will be used to determine the area for which the negative exponential dispersal kernel will be calculated
binary	A logic indicating whether the <code>occData</code> object contains presences/absences data or only the geographical coordinates of presences data (default=TRUE)
longlat	A logic indicating whether point coordinates are projected (longitude-latitude decimal degrees) or not (for plane) (default=TRUE)

Value

Returns a `RasterLayer` object of the negative exponential dispersal kernel

References

Tarek Hattab, Carol Ximena Garzon Lopez, Michael Ewald, Sandra Skowronek, Raf Aerts, Helene Horen, Boris Brasseur, Emilie Gallet-Moron, Fabien Spicher, Guillaume Decocq, Hannes Feilhauer, Olivier Honnay, Pieter Kempeneers, Sebastian Schmidlein, Ben Somers, Ruben Van De Kerchove, Duccio Rocchini and Jonathan Lenoir (2017). A unified framework to model the potential and realized distributions of invasive species within the invaded range. *Diversity and Distributions*. DOI: 10.1111/ddi.12566

Examples

```
## Not run:
library(raster)
library(sp)
envData<-getData('worldclim', var='bio', res=10)
envData<-crop(envData, extent(-8, 15, 38, 55))
envData<-envData[[1]]

#Generate randomly a SpatialPointsDataFrame containing occurrences
xy<-coordinates(envData)[sample(which(is.na(values(envData))==FALSE), 100),]
occ<-ifelse(xy[,2]>50, 0, 1)
occData<-SpatialPointsDataFrame(coords=xy, data=as.data.frame(occ),
proj4string = CRS(proj4string(envData)))

par(mfrow=c(3,3), mar=c(1.5, 1.5, 1.5, 1.5))
for (a in c(0.01, 0.03, 0.05, 0.06, 0.09, 0.1, 0.3, 0.6, 0.9)){
```

```

propagule<-iForce(occData,envData=envData,a=a,binary=TRUE,longlat=TRUE)
plot(propagule,main=paste("a = ",a))
plot(occData,col=ifelse(occData@data[,1]==1,1,0),add=TRUE,cex=0.3)}

## End(Not run)

```

pDLA *R function to compute the probability of detecting dispersal-limited absences within a network of presence-absence data*

Description

R function to compute the probability of detecting dispersal-limited absences

Usage

```
pDLA(occData,envData, longlat=TRUE,nf=5,occNative=NULL,envNative=NULL)
```

Arguments

occData	A SpatialPointsDataFrame as defined in package sp containing species data (a single species) in binary format (ones for presences, zeros for absences)
envData	An abject containing your explanatory variables. Either a SpatialPixelsDataFrame or SpatialGridDataFrame as defined in package sp or a RasterStack or RasterBrick as defined in package raster. Note that this object can contain a mixture of variables type (quantitative, factor and ordered).
longlat	A logic indicating whether point coordinates are projected (longitude-latitude decimal degrees) or not (for plane) (default=TRUE)
nf	If envData contains any factor and ordered variables, the Hill and Smith method will be used to perform an ordination of explanatory variables. nf corresponds in this case to the number of kept axes for the calulation of mahalanobis distances.
occNative	(optional) a SpatialPoint object containing the occurrences from the native range
envNative	(optional) an object of the same class as envData containing the explanatory variables from the native range

Details

This function compute the probability of detecting dispersal-limited absences within a network of presence-absence data. It's based on the idea that absences data that are environmentally close but geographically distant to presences data are more likely to be dispersal-limited absences. This function allows combining presence data from both the native and the invaded range, note that in this case the set of presence from the native range will only be added to the set of presences from the invaded range when calculating distances between absences and presences in the environmental space as it does not make sens to add presence data from the native range when calculating these distances within the geographical space.

Value

Returns a SpatialPointsDataFrame containing the probability values

References

Tarek Hattab, Carol Ximena Garzon Lopez, Michael Ewald, Sandra Skowronek, Raf Aerts, Helene Horen, Boris Brasseur, Emilie Gallet-Moron, Fabien Spicher, Guillaume Decocq, Hannes Feilhauer, Olivier Honnay, Pieter Kempeneers, Sebastian Schmidlein, Ben Somers, Ruben Van De Kerchove, Duccio Rocchini and Jonathan Lenoir (2017). A unified framework to model the potential and realized distributions of invasive species within the invaded range. Diversity and Distributions. DOI: 10.1111/ddi.12566

See Also

<<http://diarsproject.github.io/DIARS/SDM.html>>

Examples

```
## Not run:
library(raster)
library(sp)
library(virtualspecies)
library(colorRamps)

envData<-getData('worldclim', var='bio', res=10)
envData<-crop(envData,extent(-8,15,38,55))

#Generate virtual species responses with formatfunctions
my.parameters <- formatFunctions(bio1 = c(fun = "dnorm", mean = 140, sd = 40),
bio5 = c(fun = "dnorm",mean = 230, sd = 70),
bio6 = c(fun = "dnorm",mean = 10, sd = 40))

#Generate a virtual species distributions with responses to environmental variables
potential.dist <- generateSpFromFun(envData[[c(1,5,6)]], my.parameters)

#Limit a virtual species distribution to a defined area.
#It will thus generate a species which is not at the equilibrium with its environment
realized.dist<-limitDistribution(x=potential.dist$suitab.raster, area=extent(-8,15,38,48))

#Generate a random presence absence dataset from the realized distribution
# using a probability threshold of 0.5
Presence<-coordinates(realized.dist$occupied.area
)[sample(which(values(realized.dist$occupied.area)>0.5),300),]
Absence<-coordinates(realized.dist$occupied.area
)[sample(which(values(realized.dist$occupied.area)<0.5),300),]
occData<-as.data.frame(rbind(cbind(Presence,SP=rep(1,300)),cbind(Absence,SP=rep(0,300))))
coordinates(occData)<-~x+y
proj4string(occData)<-proj4string(envData)

# Calculat the probability of detecting dispersal-limited absences
probability<-pDLA(occData=occData,envData=envData[[c(1,5,6)]],longlat=TRUE)
```

```

# Display the results
par(mfrow=c(1,2),mar=c(2,2.5,2,2.5))
plot(realized.dist$occupied.area,main="Realized distribution")
plot(occData,col=ifelse(occData$SP==1,2,1),add=TRUE,pch=19,cex=0.8)
plot(potential.dist$suitab.raster,main="Potential distribution")

scatterCol<-function(x){
  x<-(x-min(x))/(max(x)-min(x))
  colorFunction <- colorRamp(matlab.like(100))
  zMatrix <- colorFunction(x)
  zColors <- rgb(zMatrix[,1], zMatrix[,2], zMatrix[,3], maxColorValue=255)
  return(zColors)}
points(probability,pch=21, col=1,bg=scatterCol(probability@data["PDLA"]),cex=1)
# Example based on occurrences from both the invaded and the native range
envData<-getData('worldclim', var='bio', res=10)
envNative<-crop(envData,extent(-90,-70,20,40))
envData<-crop(envData,extent(-8,15,38,55))
native.dist <- generateSpFromFun(envNative[[c(1,5,6)]], my.parameters)
occNative<-as.data.frame(coordinates(native.dist$suitab.raster)
[sample(which(values(native.dist$suitab.raster)>0.5),100),])
coordinates(occNative)<-~x+y
proj4string(occNative)<-proj4string(envData)
plot(native.dist,main=" Native range distribution")
plot(occNative,add=TRUE,pch=19,cex=0.8)
probability<-pDLA(occData=occData,envData=envData[[c(1,5,6)]],longlat=TRUE,
occNative=occNative,envNative=envNative[[c(1,5,6)]])

## End(Not run)

```

plotRPD	<i>R function to plot the realized versus the potential distribution using a RGB colour space</i>
---------	---

Description

R function to plot the realized versus the potential distribution using a RGB colour space

Usage

```
plotRPD(x=x,y=y,cex,xlab=xlab,ylab=ylab)
```

Arguments

x	Either a SpatialPixelsDataFrame, or a SpatialGridDataFrame as defined in package sp or a RasterLayer as defined in package raster
y	An object belonging to the same class as x
cex	A numerical value giving the amount by which plotting symbols should be magnified relative to the default

```
xlab          A title for the x axis
ylab          A title for the y axis
```

References

Tarek Hattab, Carol Ximena Garzon Lopez, Michael Ewald, Sandra Skowronek, Raf Aerts, Helene Horen, Boris Brasseur, Emilie Gallet-Moron, Fabien Spicher, Guillaume Decocq, Hannes Feilhauer, Olivier Honnay, Pieter Kempeneers, Sebastian Schmidlein, Ben Somers, Ruben Van De Kerchove, Duccio Rocchini and Jonathan Lenoir (2017). A unified framework to model the potential and realized distributions of invasive species within the invaded range. *Diversity and Distributions*. DOI: 10.1111/ddi.12566

Examples

```
## Not run:
library(raster)
library(sp)
library(virtualspecies)
library(colorRamps)

envData<-getData('worldclim', var='bio', res=10)
envData<-crop(envData,extent(-8,15,38,55))

# Generate virtual species responses with formatfunctions
my.parameters <- formatFunctions(bio1 = c(fun = "dnorm", mean = 140, sd = 40),
bio5 = c(fun = "dnorm",mean = 230, sd = 70),
bio6 = c(fun = "dnorm",mean = 10, sd = 40))
# Generate a virtual species potential distributions with responses to environmental variables
potential.dist <- generateSpFromFun(envData[[c(1,5,6)]], my.parameters)$suitab.raster

#Limit the distribution
realized.dist<-potential.dist
cell.id<-which(coordinates(realized.dist)[,2]>48)
dis.lim<-sample(seq(0,1,by=0.01),length(cell.id),replace=TRUE)
values(realized.dist)[cell.id]<-ifelse(values(realized.dist)[cell.id]>dis.lim,
values(realized.dist)[cell.id]-dis.lim,0)
plotRPD(x=potential.dist,y=realized.dist,cex=0.8,xlab="Potential distribution"
,ylab="Realized distribution")

## End(Not run)
```


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