# **Package 'ForestTools'**

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Type Package Title Analyzing Remotely Sensed Forest Data Version 0.2.5 Date 2021-09-09 Author Andrew Plowright, Jean-Romain Roussel Maintainer Andrew Plowright <andrew.plowright@alumni.ubc.ca> Description Provides tools for analyzing remotely sensed forest data, including functions for detecting treetops from canopy models, outlining tree crowns, calculating textural metrics and generating spatial statistics. **Depends** R (>= 4.0)**License** GPL (>= 3) **Encoding** UTF-8 LazyData true LinkingTo Rcpp Imports APfun, doParallel, foreach, imager, methods, parallel, plyr, progress, raster, Rcpp, rgeos, sp Suggests testthat, knitr, rmarkdown RoxygenNote 7.1.1 URL https://github.com/andrew-plowright/ForestTools BugReports https://github.com/andrew-plowright/ForestTools/issues VignetteBuilder knitr **NeedsCompilation** yes

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.calcGLCM

Calculate GLCM

#### Description

Some notes about this function: 1. Input should be a matrix 2. Shouldn't receive negative values 3. Shouldn't receive all NA values 4. Shouldn't be an empty matrix (i.e.: nrow = 0, ncol = 0) 5. 'n\_grey' shouldn't be larger than the number of unique values

#### Usage

.calcGLCM(data, n\_grey, angle, d = 1, normalize = TRUE)

#### Arguments

data	matrix. Input image
n_grey	integer. Number of grey levels used to discretize image
angle	integer. Angle at which GLCM will be calculated. Valid inputs are 0, 45, 90, or 135
d	numeric. Distance for calculating GLCM
normalize	boolean. Normalize output if TRUE

.GLCMstats

#### Description

Calculate stats for GLCM

#### Usage

.GLCMstats(data)

#### Arguments

data

matrix. GLCM computed using '.calcGLCM'

CHMde	emo
-------	-----

Canopy height model demo

#### Description

A small section of a canopy height model of a forest in British Columbia, Canada.

#### Usage

CHMdemo

#### Format

A RasterLayer

Cell values are equal to canopy height above ground (in meters)

ForestTools

Forest Tools: A package for analyzing geospatial forest data

#### Description

Forest Tools provides functions for analyzing remotely sensed forest data. Functions like vwf and mcws are applied to rasterized canopy height models, and can detect treetops and outline their respective crowns. sp\_summarise can summarize tree counts and attributes within particular areas of interest or within continuous spatial grids.

## Description

Generate textural metrics for a segmented raster using Grey-Level Co-Occurrence Matrices (GLCM). It will return a series of GLCM statistics for each segment (segs) based on an underlying singleband raster image (image) in the form of a data.frame.

#### Usage

```
glcm(
   segs,
   image,
   n_grey = 32,
   angle = 0,
   clusters = 1,
   showprog = FALSE,
   roundCoords = 4
)
```

## Arguments

segs	RasterLayer. A segmented raster. Cell values should be equal to segment numbers
image	RasterLayer. A single-band raster layer from which texture is measured
n_grey	integer. Number of grey levels the image should be quantized into
angle	integer. Angle at which GLCM will be calculated. Valid inputs are 0, 45, 90, or 135
clusters	integer. Number of clusters to use during parallel processing
showprog	logical. Display progress in terminal
roundCoords	integer. Errors in coordinate precision can trigger errors in this function. In- ternally, the coordinates are rounded to this decimal place. Default value of 4 decimals.

#### Details

The underlying C++ code for computing GLCMs and their statistics was originally written by Joel Carlson for the defunct [radiomics](https://github.com/cran/radiomics) library. It has been reused here with permission from the author.

#### Value

data.frame

4

## glcm

#### glcm0

#### References

Parmar, C., Velazquez, E.R., Leijenaar, R., Jermoumi, M., Carvalho, S., Mak, R.H., Mitra, S., Shankar, B.U., Kikinis, R., Haibe-Kains, B. and Lambin, P. (2014). *Robust radiomics feature quantification using semiautomatic volumetric segmentation. PloS one*, *9*(7)

#### Examples

```
## Not run:
# Generate raster segments
segs <- mcws(kootenayTrees, kootenayCHM, minHeight = 0.2, format = "raster")
# Get textural metrics for ortho's red band
tex <- glcm(segs, kootenayOrtho[[1]])
## End(Not run)
```

glcm0

## Create a 0 degree GLCM

## Description

Used internally by glcm()

#### Usage

glcm0(x, n\_grey, d)

#### Arguments

х	A Numeric matrix, integer values only
n_grey	Number of grey levels
d	distance from reference pixel to neighbour pixel

glcm1	35
-------	----

Create a 135 degree GLCM

## Description

Used internally by glcm()

## Usage

glcm135(x, n\_grey, d)

glcm90

## Arguments

x	A Numeric matrix, integer values only
n_grey	Number of grey levels
d	distance from reference pixel to neighbour pixel

```
glcm45
```

## Create a 45 degree GLCM

## Description

Used internally by glcm()

## Usage

glcm45(x, n\_grey, d)

## Arguments

х	A Numeric matrix, integer values only
n_grey	Number of grey levels
d	distance from reference pixel to neighbour pixel

glcm90

## Create a 90 degree GLCM

## Description

Used internally by glcm()

## Usage

glcm90(x, n\_grey, d)

## Arguments

х	A Numeric matrix, integer values only
n_grey	Number of grey levels
d	distance from reference pixel to neighbour pixel

glcm\_features GLCM Features

## Description

**GLCM** Features

#### Usage

glcm\_mean(glcm)

glcm\_variance(glcm)

glcm\_autoCorrelation(glcm)

glcm\_cProminence(glcm)

glcm\_cShade(glcm)

glcm\_cTendency(glcm)

glcm\_contrast(glcm)

glcm\_correlation(glcm)

glcm\_differenceEntropy(glcm, base = 2)

glcm\_dissimilarity(glcm)

glcm\_energy(glcm)

glcm\_entropy(glcm, base = 2)

glcm\_homogeneity1(glcm)

glcm\_homogeneity2(glcm)

glcm\_IDMN(glcm)

glcm\_IDN(glcm)

glcm\_inverseVariance(glcm)

glcm\_maxProb(glcm)

glcm\_sumAverage(glcm)

```
glcm_sumEntropy(glcm, base = 2)
```

glcm\_sumVariance(glcm)

#### Arguments

glcm	A matrix of class "glcm" produced by glcm.
base	Base of the logarithm in differenceEntropy.

#### Functions

- glcm\_mean: Mean
- glcm\_variance: Variance
- glcm\_autoCorrelation: Autocorrelation
- glcm\_cProminence: Cluster Prominence
- glcm\_cShade: Cluster Shade
- glcm\_cTendency: Cluster Tendency
- glcm\_contrast: Contrast
- glcm\_correlation: Correlation
- glcm\_differenceEntropy: Difference Entropy
- glcm\_dissimilarity: Dissimilarity
- glcm\_energy: Energy
- glcm\_entropy: Entropy
- glcm\_homogeneity1: Homogeneity
- glcm\_homogeneity2: Homogeneity 2
- glcm\_IDMN: Inverse Difference Moment (Normalized)
- glcm\_IDN: Inverse Difference (Normalized)
- glcm\_inverseVariance: Inverse Variance
- glcm\_maxProb: Maximum Probability
- glcm\_sumAverage: Sum Average
- glcm\_sumEntropy: Sum Entropy
- glcm\_sumVariance: Sum Variance

#### References

https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0102107

glcm\_img

#### Description

Get GLCM statistics for a single unsegmented image

#### Usage

glcm\_img(img, n\_grey = 32, angle = 0, d = 1, normalize = TRUE)

#### Arguments

img	matrix or raster. Input image
n_grey	integer. Number of grey levels used to discretize image
angle	integer. Angle at which GLCM will be calculated. Valid inputs are 0, 45, 90, or 135
d	numeric. Distance for calculating GLCM
normalize	boolean. Normalize output image before calculating statistics

|--|--|--|--|

#### Description

Boundaries of cut blocks within a 1.5 hectare section of forest in the Kootenay mountains, in British Columbia, Canada. Each block contains trees of different levels of maturity. Overlaps with kootenayTrees, kootenayCrowns, kootenayOrtho and kootenayCHM.

### Usage

kootenayBlocks

#### Format

A SpatialPolygonsDataFrame with the following attributes:

BlockID numerical identifier for each block

Shape\_Leng length of polygon on meters

Shape\_Area area of polygon in square meters

#### See Also

kootenayTrees kootenayCHM kootenayCrowns kootenayOrtho

kootenayCHM

## Description

A canopy height model of a 1.5 hectare section of forest in the Kootenay mountains, in British Columbia, Canada.

## Usage

kootenayCHM

#### Format

A RasterLayer Cell values are equal to canopy height above ground (in meters)

#### Source

Data acquired from a photogrammetric drone survey performed by Spire Aerobotics on June 16th, 2016.

## See Also

kootenayTrees kootenayBlocks kootenayCrowns kootenayOrtho

|--|

## Description

Outlines of tree crowns corresponding to the kootenayTrees treetops. Generated using mcws.

#### Usage

kootenayCrowns

#### Format

A SpatialPolygonsDataFrame with the following attributes:

**height** height of the tree's apex, in meters above ground. Inherited from kootenayTrees. **winRadius** radius of the moving window at the treetop's location. Inherited from kootenayTrees. **crownArea** area of crown outline in square meters

#### See Also

kootenayTrees kootenayCHM kootenayBlocks kootenayOrtho

kootenay0rtho

#### Description

An orthomosaic of a 1.5 hectare section of forest in the Kootenay mountains, in British Columbia, Canada.

## Usage

kootenayOrtho

#### Format

A RasterLayer

Cell values are equal to canopy height above ground (in meters)

#### Source

Data acquired from a photogrammetric drone survey performed by Spire Aerobotics on June 16th, 2016.

### See Also

kootenayTrees kootenayBlocks kootenayCrowns kootenayCHM

kootenayTrees

Kootenay forest - Dominant trees over 2 m

#### Description

Dominant trees from a 1.5 hectare section of forest in the Kootenay mountains, in British Columbia, Canada. Trees were detected by applying the vwf function to the kootenayCHM raster dataset. Only trees over 2 m above ground were detected.

#### Usage

kootenayTrees

#### Format

A SpatialPointsDataFrame with the following attributes:

height height of the tree's apex, in meters above ground

winRadius radius of the moving window (see vwf) at the treetop's location

mcws

#### See Also

kootenayCHM kootenayBlocks kootenayCrowns kootenayOrtho

mcws

Marker-Controlled Watershed Segmentation

#### Description

Implements the watershed function to segment (i.e.: outline) crowns from a canopy height model. Segmentation is guided by the point locations of treetops, typically detected using the vwf function. See Meyer & Beucher (1990) for details on watershed segmentation.

#### Usage

```
mcws(
   treetops,
   CHM,
   minHeight = 0,
   format = "raster",
   OSGeoPath = NULL,
   verbose = FALSE
)
```

#### Arguments

treetops	SpatialPointsDataFrame. The point locations of treetops. The function will generally produce a number of crown segments equal to the number of treetops.
СНМ	Canopy height model in raster format. Should be the same that was used to create the input for treetops.
minHeight	numeric. The minimum height value for a CHM pixel to be considered as part of a crown segment. All CHM pixels beneath this value will be masked out. Note that this value should be lower than the minimum height of treetops.
format	string. Format of the function's output. Can be set to either 'raster' or 'poly- gons'.
OSGeoPath	character. Optional path to the OSGeo4W installation directory. If both OS-Geo4W and Python are installed, this will enable the function to use a faster algorithm for producing polygonal crown outlines (see Details below).
verbose	logical. Print processing progress to console.

#### Details

This function can return a crown map as either a raster or a SpatialPolygonsDataFrame, as defined using the format argument. For most analytical purposes, it is preferable to have crown outlines as polygons. However, polygonal crown maps take up significantly more disk space, and take longer to process. It is advisable to run this function using a raster output first, in order to check its results and adjust parameters.

#### mcws

Using the polygons provides the added benefit of transferring treetop attributes (such as *height*) to the newly created polygons. The area of each crown will also automatically be calculated and added to the polygons' data under the *crownArea* field. Furthermore, "orphaned" segments (i.e.: outlines without an associated treetop) will be removed when format is set to 'polygons'.

By default, polygonal crown outlines are produced internally using the the rasterToPolygons function from the raster package. This function is problematic due to it being 1) very slow and 2) leaking memory when applied to multiple datasets. An alternative is provided for users who've installed OSGeo4W and Python. By setting the OSGeoPath path to the OSGeo4W installation directory (usually 'C:\OSGeo4W64'), the function will use the *gdal\_polygonize.py* GDAL utility to generate polygonal crown outlines instead.

#### Value

Depending on the argument set with format, this function will return a map of outlined crowns as either a RasterLayer (see raster), in which distinct crowns are given a unique cell value, or a SpatialPolygonsDataFrame, in which each crown is represented by a polygon.

#### References

Meyer, F., & Beucher, S. (1990). Morphological segmentation. *Journal of visual communication* and image representation, 1(1), 21-46.

#### See Also

vwf sp\_summarise watershed

OSGeo4W download page: https://trac.osgeo.org/osgeo4w/

#### Examples

```
## Not run:
# Use variable window filter to detect treetops in demo canopy height model
ttops <- vwf(CHMdemo, winFun = function(x){x * 0.06 + 0.5}, minHeight = 2)
# Set minimum tree crown height (should be LOWER than minimum treetop height)
minCrwnHgt <- 1
# Use 'mcws' to outline tree crowns
segs <- mcws(ttops, CHMdemo, minCrwnHgt)
## End(Not run)
```

quesnelBlocks

#### Description

Boundaries of cut blocks within a 125 hectare section of forest in the Quesnel Timber Supply Area, in British Columbia, Canada. Each block contains trees of different levels of maturity. Overlaps with quesnelTrees and quesnelCHM.

#### Usage

quesnelBlocks

#### Format

A SpatialPolygonsDataFrame with the following attributes:

BlockID numerical identifier for each block

Shape\_Leng length of polygon on meters

Shape\_Area area of polygon in square meters

#### See Also

quesnelTrees quesnelCHM

quesnelCHM

Quesnel forest - Canopy height model

#### Description

A canopy height model of a 125 hectare section of forest in the Quesnel Timber Supply Area, in British Columbia, Canada.

#### Usage

quesnelCHM

#### Format

A RasterLayer

Cell values are equal to canopy height above ground (in meters)

#### Source

Data acquired from a photogrammetric drone survey performed by Spire Aerobotics on September 15th, 2016.

#### quesnelTrees

#### See Also

quesnelTrees quesnelBlocks

quesnelTrees

Quesnel forest - Dominant trees over 2 m

#### Description

Dominant trees from a 125 hectare section of forest in the Quesnel Timber Supply Area, in British Columbia, Canada. Trees were detected by applying the vwf function to the quesnelCHM raster dataset. Only trees over 2 m above ground were detected.

### Usage

quesnelTrees

#### Format

A SpatialPointsDataFrame with the following attributes:

height height of the tree's apex, in meters above ground

winRadius radius of the moving window (see vwf) at the treetop's location

#### See Also

quesnelCHM quesnelBlocks

sp\_summarise Spatial Summarization

#### Description

Summarization tool for calculating tree counts and statistics within various spatial units.

#### Usage

```
sp_summarise(
  trees,
  areas = NULL,
  grid = NULL,
  variables = NULL,
  statFuns = NULL
)
```

#### Arguments

trees	SpatialPointsDataFrame or SpatialPolygonsDataFrame. The locations of a set of trees, typically detected from a canopy height model using vwf. Tree attributes, such as height or crown size, should be stored within this object's @data slot. Tree crowns delineated using mcws can also be used.
areas	SpatialPolygonsDataFrame. An optional set of polygons corresponding to areas of interest. Tree counts and statistics will be returned for each area.
grid	RasterLayer (see raster) or numeric. An alternative to the areas argument. Us- ing grid will compute tree counts and statistics within the cells of a spatial grid. Grid size and placement can be defined by inputting a raster object. A single numeric value can also be used, in which case the function will generate a grid with a cell size equal to this value.
variables	character. The names of tree attribute variables (stored in the trees@data slot). In addition to tree counts, the function will compute statistics for each of these variables. Only numeric variables are accepted.
statFuns	list. A named list of custom functions that are used to compute tree attribute statistics. If none are provided, default statistics are mean, median, standard deviation, minimum and maximum. Note that each element of the list should have a name that describes the statistics generated by the function. See below for details on defining custom functions.

#### Details

Input trees can either be point locations (SpatialPointsDataFrame) or crown outlines (SpatialPolygonsDataFrame). If crown outlines (or other polygons) are inputted, they will be partitioned between spatial units according to their geographic centroids.

In addition to tree counts, statistics for the trees' attributes can also be generated. These attributes should be defined within the @data slot of the input. Only numeric variables are accepted.

By default, the statistics generated for each attribute will be its mean, median, standard deviation, minimum and maximum. However, custom functions can also be used with the statFuns argument. This should be a named list of functions, wherein each list element is given a name to represent the statistic computed by the function.

For example: list(qunt98 = function(x, ...) quantile(x, c(.98), na.rm = TRUE))

Furthermore, custom functions should:

- Be able to accept numeric vectors.
- Be able to handle NA values.
- Have an ellipsis (three dots) in their arguments: function(x, ...)
- Return a single numeric value.

#### Value

Tree count and, if any variables are supplied, tree attribute statistics. If no areas or grid is supplied, the tree count and statistics are computed for the entire trees dataset, and returned as a 'data.frame' object. If areas are defined, an identical SpatialPolygonsDataFrame will be returned,

vwf

with all computed statistics appended to the object's @data slot. If a grid is defined, tree count will be returned as a RasterLayer, with cell values equal to the number of trees in each cell. If a grid and variables are defined, a RasterBrick (see brick) will be returned instead, with tree count and attribute statistics stored as stacked layers.

#### See Also

vwf mcws

#### Examples

```
# Load sample data
library(ForestTools)
library(sp)
data("kootenayTrees", "kootenayBlocks", "kootenayCrowns")
# Get total tree count
sp_summarise(kootenayTrees)
# Get total tree count, tree height and crown area statistics
sp_summarise(kootenayCrowns, variables = c("height", "crownArea"))
# Get tree count, height statistics for specific areas of interest
areaStats <- sp_summarise(kootenayTrees, areas = kootenayBlocks, variables = "height")
# Plot according to tree count
plot(areaStats, col = heat.colors(3)[order(areaStats$TreeCount)])
# Get tree count and height statistics for a 20 x 20 m spatial grid
gridStats <- sp_summarise(kootenayTrees, grid = 20, variables = "height")</pre>
# Plot gridded tree count and statistics
plot(gridStats$TreeCount)
plot(gridStats$heightMax)
```

vwf

Variable Window Filter

#### Description

Implements the variable window filter algorithm (Popescu & Wynne, 2004) for detecting treetops from a canopy height model.

#### Usage

vwf( CHM, winFun,

```
minHeight = NULL,
maxWinDiameter = 99,
minWinNeib = "queen",
verbose = FALSE
)
```

## Arguments

СНМ	Canopy height model. Either in raster format, or a path directing to a raster file. A character vector of multiple paths directing to a tiled raster dataset can also be used.
winFun	function. The function that determines the size of the window at any given location on the canopy. It should take the value of a given CHM pixel as its only argument, and return the desired *radius* of the circular search window when centered on that pixel. Size of the window is in map units.
minHeight	numeric. The minimum height value for a CHM pixel to be considered as a po- tential treetop. All CHM pixels beneath this value will be masked out.
maxWinDiameter	numeric. Sets a cap on the maximum window diameter (in cells). If an improp- erly calibrated function is set for winFun, it may produce overly large windows that would perform poorly and significantly slow processing time. This setting can be disabled by setting to NULL.
minWinNeib	character. Define whether the smallest possible search window $(3x3)$ should use a queen or a rook neighborhood.
verbose	logical. Print progress to console if set to TRUE.

#### Details

This function uses the resolution of the raster to figure out how many cells the window needs to cover. This means that the raster value (representing height above ground) and the map unit (represented by the raster's resolution), need to be in the \_same unit\_. This can cause issues if the raster is in lat/long, whereby its resolution is in decimal degrees.

#### Value

SpatialPointsDataFrame. The point locations of detected treetops. The object contains two fields in its data table: *height* is the height of the tree, as extracted from the CHM, and *winRadius* is the radius of the search window when the treetop was detected. Note that *winRadius* does not necessarily correspond to the radius of the tree's crown.

#### References

Popescu, S. C., & Wynne, R. H. (2004). Seeing the trees in the forest. *Photogrammetric Engineering & Remote Sensing*, 70(5), 589-604.

#### See Also

mcws sp\_summarise

vwf

## Examples

```
# Set function for determining variable window radius winFunction <- function(x){x \star 0.06 + 0.5}
```

# Set minimum tree height (treetops below this height will not be detected) minHgt <- 2

# Detect treetops in demo canopy height model
ttops <- vwf(CHMdemo, winFunction, minHgt)</pre>

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