

# Post-Hurricane Damage Severity Classification at the Individual Tree Level Using Terrestrial Laser Scanning and Deep Learning

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## rTLsDeep Package



Figure S1. Cover page of the rTLsDeep package

rTLsDeep: An R Package for post-hurricane damage severity classification at the individual tree level using terrestrial laser scanning and deep learning.

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The rTLsDeep package provides options for i) rotating and deriving 2D images from TLS 3D point clouds, ii) calibrating and validating convolutional neural network (CNN) architectures and iii) predicting post-hurricane damage severity at the individual tree level

## Getting Started

### Install R, Git and Rtools40

i) R ( $\geq 4.0.0$ ): <https://www.r-project.org/>

ii) Git: <https://git-scm.com/>

iii) tensorflow (python environment): <https://doi.org/10.5281/zenodo.3929709>

### rTLsDeep installation

```
# The CRAN version:
```

```
install.packages("rTLsDeep")
```

```
# The development version:
```

```
#install.packages("remotes")
```

```
library(remotes)
```

```
install_github("https://github.com/carlos-alberto-silva/rTLsDeep", dependencies = TRUE)
```

## Getting Started

### Loading rTLsDeep and other required packages

```
# get pacman
```

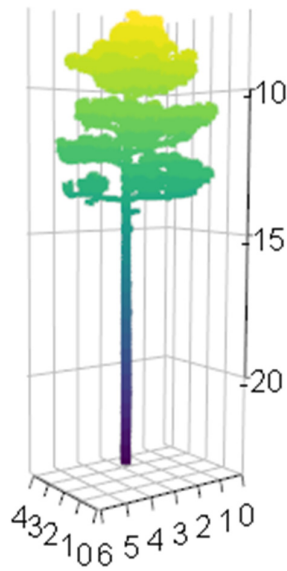
```
install.packages("pacman")
```

```
#load pacman and all packages
```

```
library(pacman)
```

```
p_load(rTLsDeep,lidR,rgl,ggplot2,rgl,keras,reticulate,compiler)
```

### TLS data processing



**Figure S2.** Example of TLS-derived 3d point cloud

### Loading and visualizing TLS dataset

```
# Path to las file
lasfile <- system.file("extdata", "tree_c1.laz", package="rTLsDeep")

# Reading las file
las<-readLAS(lasfile)

# plotting las file in 3D
plot(las, bg="white")
rgl::axes3d(c("x+", "y-", "z-"), col="black")
rgl::grid3d(side=c('x+', 'y-', 'z'), col="gray")
```



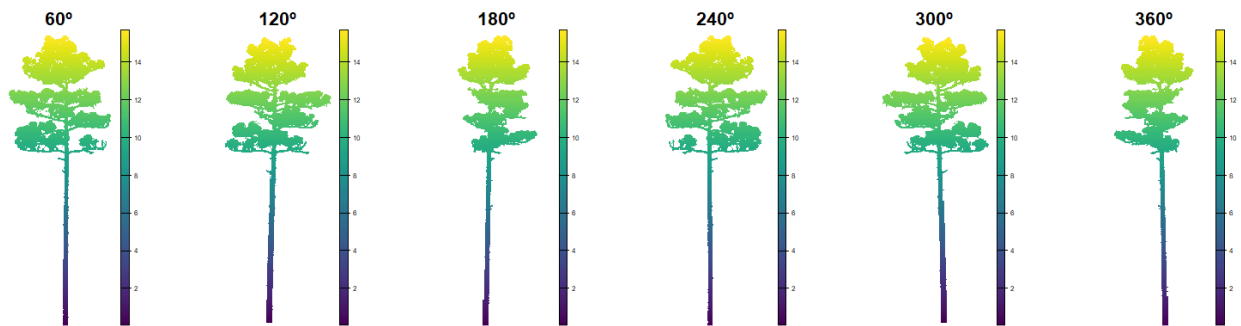
**Figure S3.** Illustration of the TLS-derived 3d point cloud

## Rotating TLS-derived 3d point cloud

```
# Rotating around the x-axis
las<-tlsrotate3d(las,theta=120, by="x", scale=TRUE)

# Rotating around the y-axis
las<-tlsrotate3d(las,theta=120, by="y", scale=TRUE)

# Rotating around the z-axis
las<-tlsrotate3d(las,theta=120, by="z", scale=TRUE)
```



**Figure S4.** Illustration of 3d point cloud rotation

## Capturing 2D grid snapshot

```
# Set output dir for downloading the example dataset files
outdir=getwd()

# downloading zip file
download.file("https://github.com/carlos-alberto-
  silva/rTLsDeep/tree/main/readme/laz_files.zip",destfile=file.path(outdir, "laz_files.zip"))

# unzip file
unzip(file.path(outdir,"laz_files.zip"))

# Reading las file for each post-hurricane individual tree-level damage classes
tree_c1<-readLAS(paste0(outdir,"//Tree_c1.laz"))
tree_c2<-readLAS(paste0(outdir,"//Tree_c2.laz"))
tree_c3<-readLAS(paste0(outdir,"//Tree_c3.laz"))
tree_c4<-readLAS(paste0(outdir,"//Tree_c4.laz"))
tree_c5<-readLAS(paste0(outdir,"//Tree_c5.laz"))
tree_c6<-readLAS(paste0(outdir,"//Tree_c6.laz"))

# Defining the func parameter
func = ~list(Z = max(Z)) # plot by height

# computing 2D grid snapshot
```

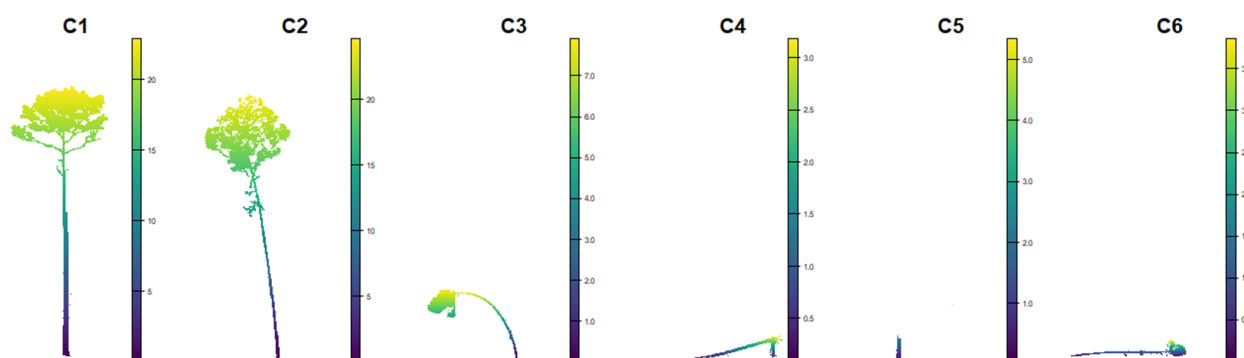
```

gtree_c1<-getTLS2D(tree_c1, res=0.05, by="xz", func = func, scale=TRUE)
gtree_c2<-getTLS2D(tree_c2, res=0.05, by="xz", func = func, scale=TRUE)
gtree_c3<-getTLS2D(tree_c3, res=0.05, by="xz", func = func, scale=TRUE)
gtree_c4<-getTLS2D(tree_c4, res=0.05, by="xz", func = func, scale=TRUE)
gtree_c5<-getTLS2D(tree_c5, res=0.05, by="xz", func = func, scale=TRUE)
gtree_c6<-getTLS2D(tree_c6, res=0.05, by="xz", func = func, scale=TRUE)

# Visualizing 2D grid snapshot
par(mfrow=c(2,3))
plot(gtree_c1, col=viridis::viridis(100),axes=FALSE, xlab="", ylab="", ylim=c(0,30), main="C1",cex=2)
plot(gtree_c2, col=viridis::viridis(100),axes=FALSE, xlab="", ylab="", ylim=c(0,30), main="C2",cex=2)
plot(gtree_c3, col=viridis::viridis(100),axes=FALSE, xlab="", ylab="", ylim=c(0,30), main="C3",cex=2)
plot(gtree_c4, col=viridis::viridis(100),axes=FALSE, xlab="", ylab="", ylim=c(0,30), main="C4",cex=2)
plot(gtree_c5, col=viridis::viridis(100),axes=FALSE, xlab="", ylab="", ylim=c(0,30), main="C5",cex=2)
plot(gtree_c6, col=viridis::viridis(100),axes=FALSE, xlab="", ylab="", ylim=c(0,30), main="C6",cex=2)

#Exporting 2D grid snapshot as tiff file
tiff("gtree_c1.tiff", units="in", width=5, height=5, res=300)
plot(gtree_c1, col=viridis::viridis(100),axes=FALSE, xlab="", ylab="", ylim=c(0,30), main="C1",cex=2)
dev.off()

```



**Figure S5.** Illustration of the damage severity classes from TLS-derived 3d point cloud

## Post-hurricane individual tree-level damage using deep learning

### Selecting deep learning model properties

```

# Set directory to tensorflow (python environment)
# This is required if running deep learning local computer with GPU
# Guide to install here: https://doi.org/10.5281/zenodo.3929709
tensorflow_dir = '/apps/tensorflow/2.6.0'

```

```
# define model type
#model_type = "simple"
model_type = "vgg"
#model_type = "inception"
#model_type = "resnet"
#model_type = "densenet"
#model_type = "efficientnet"

# path to image folders - black
train_image_files_path <- getwd() # update the path for training datasets
test_image_files_path <- getwd() # update the path for testing datasets

# Image and model properties
img_width <- 256
img_height <- 256
class_list_train = unique(list.files(train_image_files_path))
class_list_test = unique(list.files(test_image_files_path))
lr_rate = 0.0001
target_size <- c(img_width, img_height)
channels <- 4
batch_size = 8L
epochs = 20L

# get model
model = get_dl_model(model_type=model_type,
  img_width=img_width,
  img_height=img_height,
  lr_rate = lr_rate,
  tensorflow_dir = tensorflow_dir,
  class_list = class_list_train)
```

## Model calibration

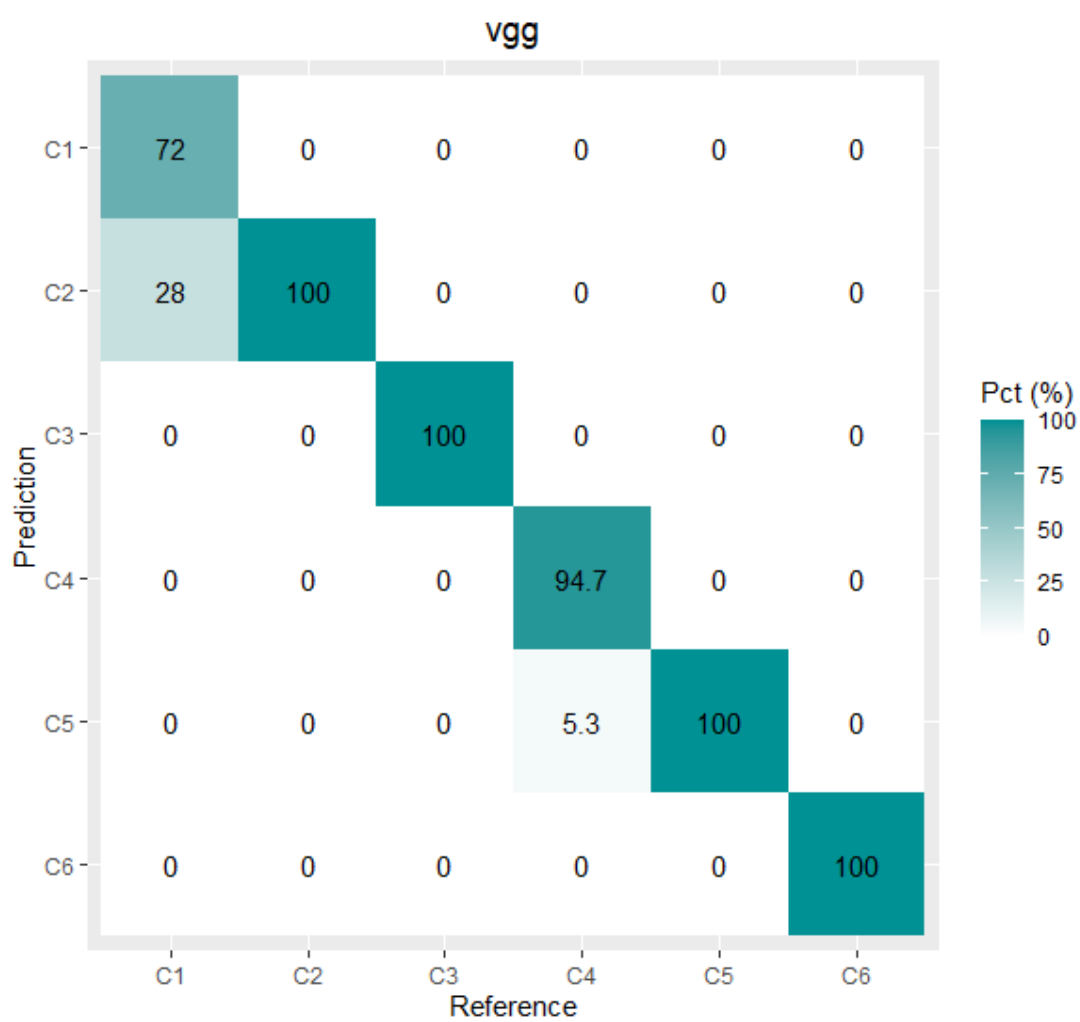
```
weights_fname = fit_dl_model(model = model,
  train_input_path = train_image_files_path,
  test_input_path = test_image_files_path,
  target_size = target_size,
  batch_size = batch_size,
  class_list = class_list_train,
  epochs = epochs,
  lr_rate = lr_rate)
```

## Predicting post-hurricane damage at the tree-level

```
tree_damage<-predict_treedamage(model = model,  
    input_file_path = test_image_files_path,  
    weights = weights,  
    target_size = c(256,256),  
    class_list=class_list_test,  
    batch_size = batch_size)
```

## Confusion matrix

```
# Get damage classes for validation datasets  
test_classes<-get_test_classes(file_path=test_image_files_path)  
  
# Calculate confusion matrix  
cm = confmatrix_treedamage(predict_class = tree_damage,  
    test_classes=test_classes,  
    class_list = class_list_test)  
  
# Plot confusion matrix  
gcmplot_vgg<-gcmplot(cm,  
    colors=c(low="white", high="#009194"),  
    title="densenet")
```



**Figure S6.** Confusion matrix for the post-hurricane damage severity classification