

# Flood change detection

Yrneh Ulloa, Seerin Herzsprung

15.10.2021

## 1. Set working directory

```
# Set the working directory ("wd") and assign it to an object (here, "w")
w <- setwd(".\\your\\path")

# Check the location of the working directory
getwd()
```

## 2. Load and install packages

```
ipak <- function(pkg){
  newpkg <- pkg[!(pkg %in% installed.packages()[, "Package"])]
  if (length(newpkg))
    install.packages(newpkg, dependencies = TRUE)
  sapply(pkg, require, character.only = TRUE)
}

packages <- c("rgdal", "raster", "sp", "RColorBrewer", "RStoolbox", "sf")
ipak(packages)
```

```
## Lade nötiges Paket: rgdal
```

```
## Lade nötiges Paket: sp
```

```
## Please note that rgdal will be retired by the end of 2023,
## plan transition to sf/stars/terra functions using GDAL and PROJ
## at your earliest convenience.
##
## rgdal: version: 1.5-27, (SVN revision 1148)
## Geospatial Data Abstraction Library extensions to R successfully loaded
## Loaded GDAL runtime: GDAL 3.2.1, released 2020/12/29
## Path to GDAL shared files: C:/Program Files/R/R-4.1.1/library/rgdal/gdal
## GDAL binary built with GEOS: TRUE
## Loaded PROJ runtime: Rel. 7.2.1, January 1st, 2021, [PJ_VERSION: 721]
## Path to PROJ shared files: C:/Program Files/R/R-4.1.1/library/rgdal/proj
## PROJ CDN enabled: FALSE
## Linking to sp version:1.4-5
## To mute warnings of possible GDAL/OSR exportToProj4() degradation,
## use options("rgdal_show_exportToProj4_warnings"="none") before loading sp or rgdal.
## Overwritten PROJ_LIB was C:/Program Files/R/R-4.1.1/library/rgdal/proj
```

```
## Lade nötiges Paket: raster
```

```
## Lade nötiges Paket: RColorBrewer
```

```
## Lade nötiges Paket: RStoolbox
```

```
## Lade nötiges Paket: sf
```

```
## Linking to GEOS 3.9.1, GDAL 3.2.1, PROJ 7.2.1
```

Another way of installing the packages one by one, is the following:

```
# Install packages function when it is not necessary to specify the folder (because this path was saved in the Global options of R)  
install.packages("rgdal")
```

```
# Install multiple packages simultaneously  
install.packages(c("sp", "raster")) #to install more than one package at the same time
```

```
# Load the packages  
library(sp)  
library(rgdal)  
library(raster)  
library(RStoolbox)  
library(RColorBrewer)
```

### 3. Create change data

For detecting change, you can make classes for the observed change in the landscape. For example:

- 1 : agriculture - agriculture
- 2 : forest - forest
- 3 : river - sediment
- 4 : urban - urban
- 5 : urban - sediment
- 6 : river - river

In QGIS, create at least 10-15 polygons per land class. These polygons correspond to the change data that will be used for the supervised classification.

Use the Sentinel-2 RasterStack image

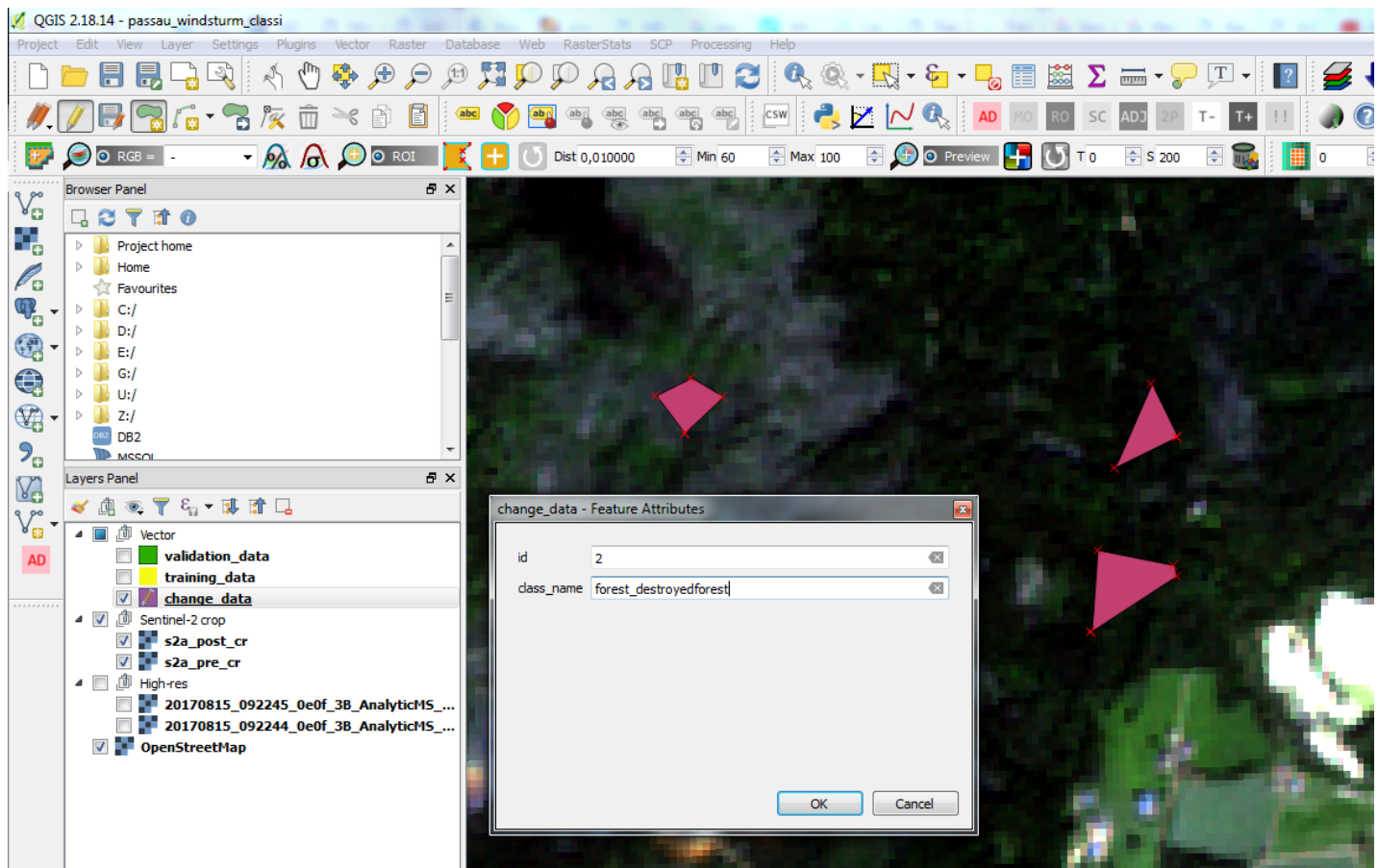


Figure 1. Creation of change data.

## 4. Change detection: Classification of rasters

### Read raster data

```
s2_pre_cr <- stack("data\\s2_change_rasters\\juneCrop_32N.tif")
s2_post_cr <- stack("data\\s2_change_rasters\\julyCrop_32N.tif")
```

Merge the before and after Sentinel-2 images for the classification

```
s2_pre_cr <- resample(s2_pre_cr, s2_post_cr)
s2_change <- stack(s2_pre_cr, s2_post_cr)
s2_change
```

```
## class      : RasterStack
## dimensions : 357, 512, 182784, 8  (nrow, ncol, ncell, nlayers)
## resolution : 19.9, 19.9  (x, y)
## extent     : 346382.5, 356571.3, 5587894, 5594998  (xmin, xmax, ymin, ymax)
## crs        : +proj=utm +zone=32 +datum=WGS84 +units=m +no_defs
## names      : juneCrop_32N.1, juneCrop_32N.2, juneCrop_32N.3, juneCrop_32N.4, julyCrop_32N.1,
julyCrop_32N.2, julyCrop_32N.3, julyCrop_32N.4
## min values :      77.35617,      192.66094,      39.42276,      316.78022,      34.02091,
83.35627,      40.24007,      108.95861
## max values :      5711.047,      6187.704,      6608.060,      6988.164,      9506.598,
8726.718,      9487.597,      9001.882
```

## Read vector data with land cover change classes

In this example, we created a multipolygon shapefile with 6 classes that corresponds to the land covers identifiable on the raster image:

- 1 : agriculture - agriculture
- 2 : forest - forest
- 3 : river - sediment
- 4 : urban - urban
- 5 : urban - sediment
- 6 : river - river

Import the vector file with these landcover change classes

```
# Import change samples
change <- readOGR("data\\vector\\wk3_change_data.shp")

# Since the projection did not match, I made a reprojection here.
change_32N <- spTransform(change, crs(s2_change))
```

Plot the change data on top of the raster subset.

```
# Plot the raster with both vector files on top
# First, add the raster
plotRGB(s2_pre_cr, r=3, g=2, b=1, axes=FALSE, stretch="lin")

# Second, add the vectors
plot(change_32N, col="red", add=TRUE)
```



Figure 2. Change samples on Sentinel RasterStack

### Classification with `superClass()` from the RStoolbox Package

```
flood_change <- superClass(img = s2_change, nSamples=100, trainData = change_32N, responseCol =  
"class_name")
```

```
# Display classification results  
flood_change
```

```
## superClass results
## ***** Validation *****
## $validation
## [1] "No independent validation was performed!"
##
## ***** Map *****
## $map
## class      : RasterLayer
## dimensions : 357, 512, 182784 (nrow, ncol, ncell)
## resolution : 19.9, 19.9 (x, y)
## extent     : 346382.5, 356571.3, 5587894, 5594998 (xmin, xmax, ymin, ymax)
## crs        : +proj=utm +zone=32 +datum=WGS84 +units=m +no_defs
## source     : memory
## names      : class_name
## values     : 1, 6 (min, max)
## attributes :
##      ID      value
## from: 1      agri_agri
## to  : 6      urban_urban
```

Since, the validation data was not included, therefore, no test of the model could be performed. Create a validation data, include it in the model and run again the classifier. Share the results of this step with your colleagues.

Plot your change detection map

```
plot(flood_change$map, breaks = c(0, 1, 2, 3, 4, 5, 6), col = c("darkolivegreen", "tomato", "yellowgreen", "tan", "blue2", "cyan"), main = 'Change Detection Classification')
```

## Change Detection Classification

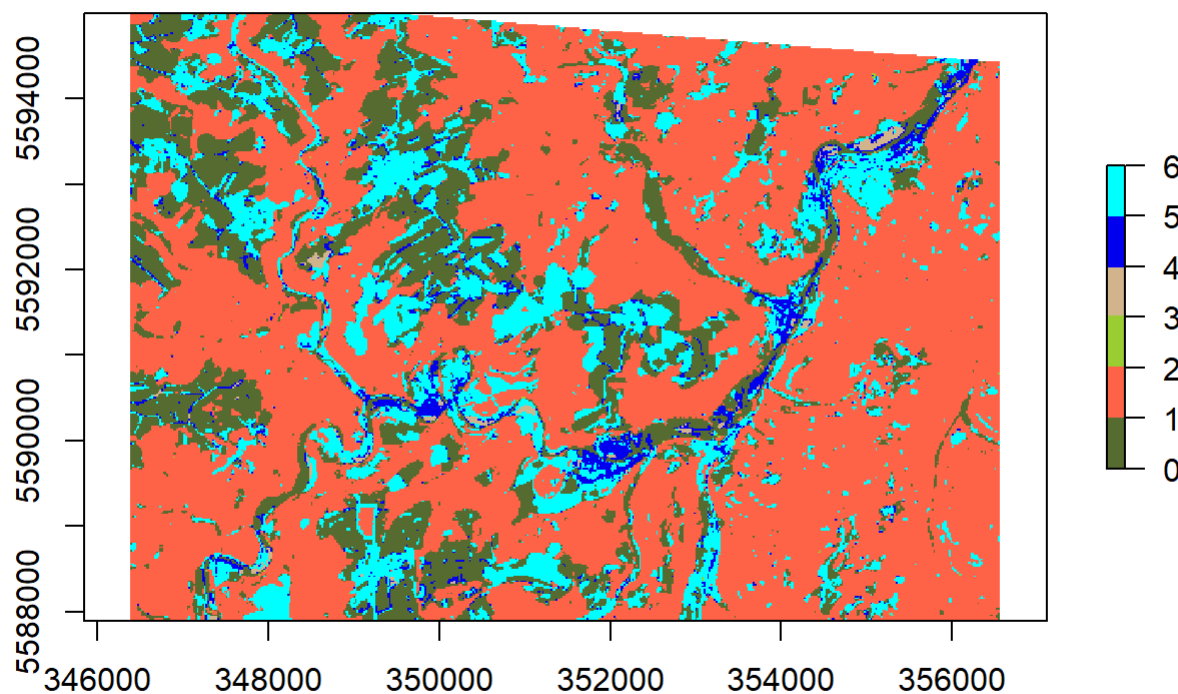


Figure 3. Change detection map

## 5. Landscape Statistics

Generate statistics of the area that was classified to quantify the damages

```
flood_change.freq <- freq(flood_change$map, useNA= "no")
```

```
change.freq <- flood_change.freq[, "count"]*10^2*1e-06
```

```
barplot(change.freq, main = "Area (km2) of landcover change in our study area",
        col= c("darkolivegreen", "tomato", "yellowgreen", "tan", "blue2", "cyan"),
        names.arg = c("agriculture", "forest", "river", "sedimented river", "sedimented urban",
"urban"))
```

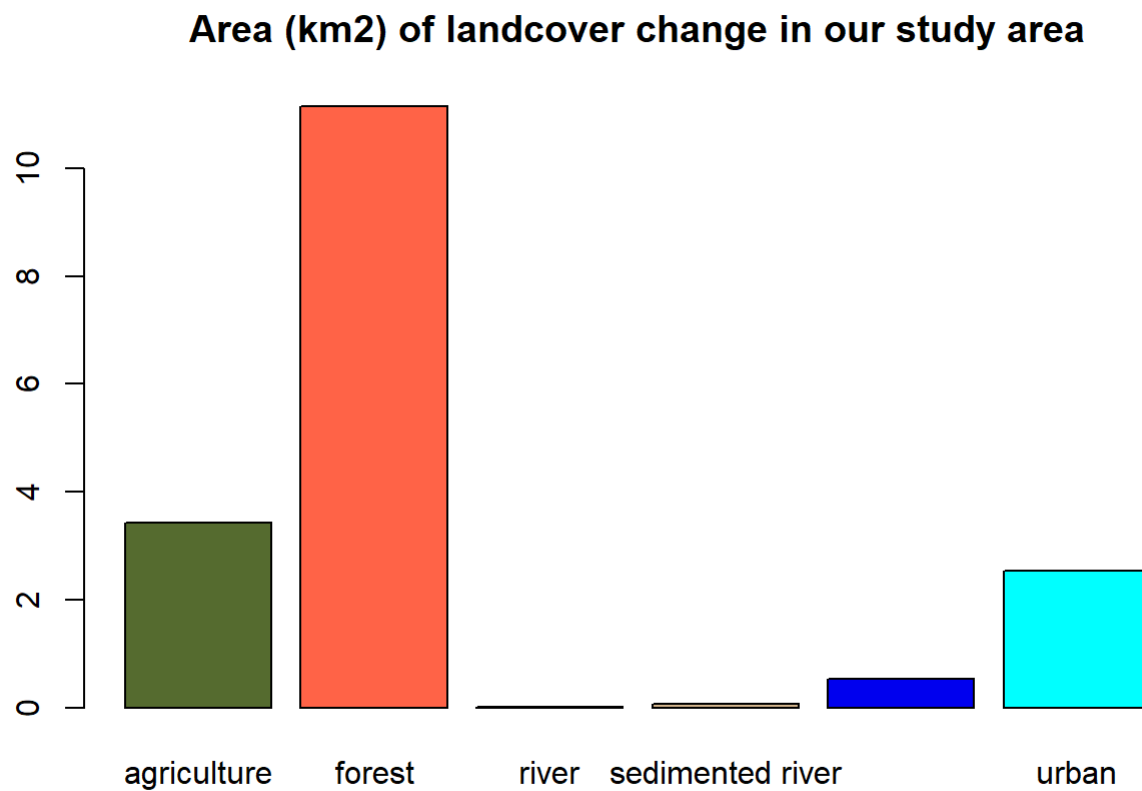


Figure 4. Area per land cover class

```
map_classi <- flood_change$map  
  
writeRaster(map_classi, datatype="FLT4S", filename = "data/wk3_results\\change_detection_classification.tif", format = "GTiff", overwrite=TRUE)
```