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Kompleksowa analiza wylesiania w krajach tropikalnych - bezpośrednie czynniki wylesiania, emisje dwutlenku węgla i równowaga wartości usług ekosystemów

A comprehensive study on deforestation in the tropics - direct deforestation drivers, carbon emissions and ecosystem service value balance

Master's Thesis
on the course of - Forestry

Thesis written under the supervision of
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Research Domain II - Climate Climate Impacts & Vulnerabilities

Potsdam, 2019

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1 Introduction

1.1 Tropical forest

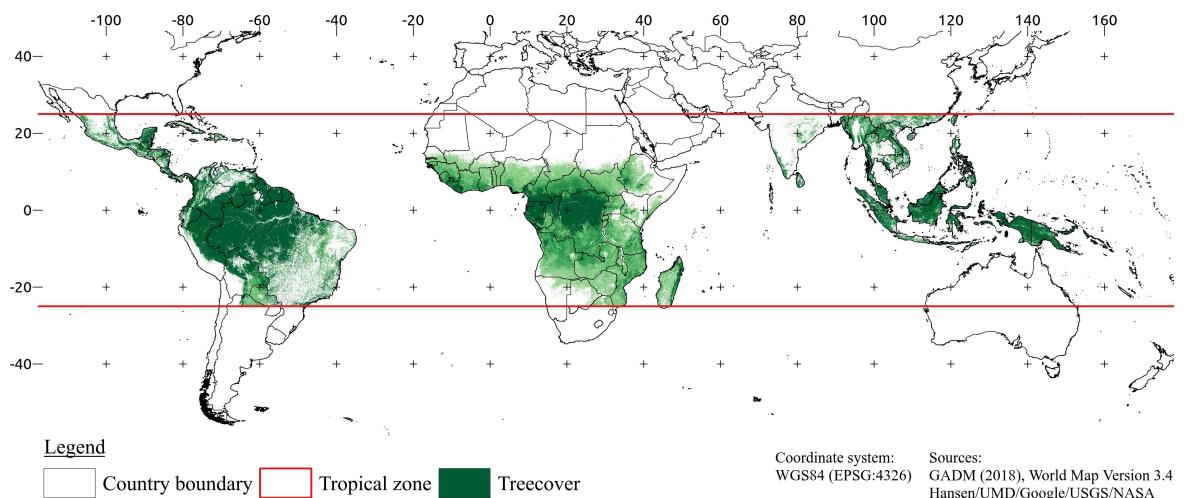


Figure 1.1. Geographic tropical zone framed red and the tropical forest

1.1.1 Current state

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1.1.2 Contribution to climate

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1.1.3 Forest definitions

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1.2 Deforestation

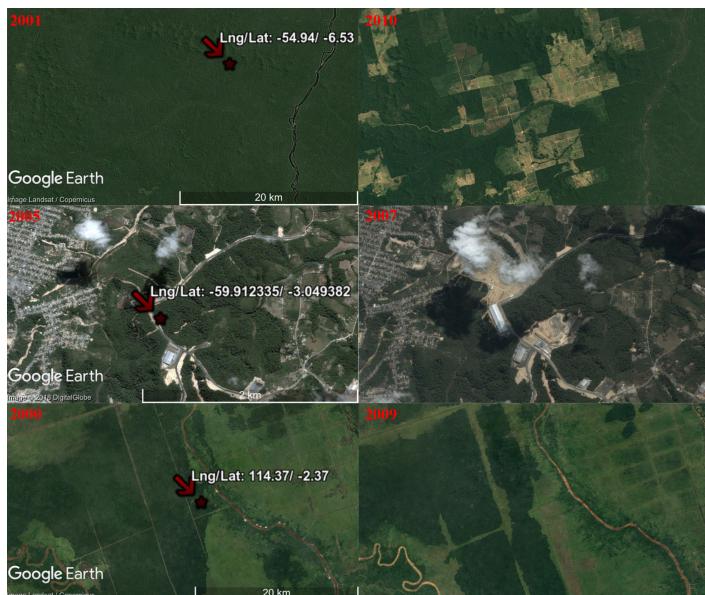


Figure 1.2. Upper Brazil agriculture, middle Brazil urbanization, lower Indonesia large scale palm oil plantations

1.2.1 Land use and land cover change

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1.2.2 Drivers of deforestation

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1.2.3 Emissions trough deforestation

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1.2.4 Removal of AGB

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1.2.5 Soil organic carbon change and soil dynamics

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1.3 Ecosystem services

1.3.1 Ecosystem service values

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1.3.2 Research objective and questions

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2 Data and methods

2.1 Data

2.1.1 Spatial data

2.1.1.1 Global Forest Change

Global Forest Change (GFC) 2000-2012 Version 1.0 is the first high-resolution dataset that provides a comprehensive view of the annual global forest cover change between 2000 and 2012 [Hansen et al. 2013; Li et al. 2017]. The initial GFC dataset released by Hansen et al. is extended by recent releases which encompass the annual forest cover changes between 2000-2013, 2000-2014, 2000-2015 and 2000-2016, respectively. All versions of this dataset have in common, that they are derived from growing season imagery captured by the remote sensing satellite Landsat 7 Enhanced Thematic Mapper Plus (ETM+) enhanced by band metrics of other sensors like Quickbird imagery, existing percent tree cover layers from Landsat data, and global Moderate Resolution Imaging Spectroradiometer (MODIS) percent tree cover [Hansen et al. 2013]. On the satellite imagery, a time-series spectral metrics analysis is applied to gather the global forest extent at 2000 as well as the annual forest loss and the accumulated gain for the period 2001 till 2012. Hence, GFC comprises three independent data layers tree cover, annually forest loss and forest gain divided into 10x10 degree tiles by the geodetic coordinate system World Geodetic System 1984 (WGS84) (EPSG:4326) with a spatial resolution of 1 arc-second per pixel (approximately 900 Km² or 30x30 m). Furthermore, across the provided Geo-Tiff (GTiff) layers the pixel data is coded in unsigned 8-bit integers. Hansen et al. defined trees as all vegetation taller than 5 meters for their study. For each pixel covered by trees, a canopy density ranging from 0 to 100 % is computed. Forest loss is defined as a stand displacement disturbance leading from a forest state to a non-forest state (e.g. canopy density >50 % to 0). Tree cover gain is defined as the inverse of loss where the canopy density must exceed 50 % to get recognized.

Hansen et al. reports as an accuracy assessment of tree cover loss a producers accuracy of approximately 83 % for the tropical region. The mapped tree cover gain is probably an

underestimation of the true gain with a producers accuracy of 48 % and a user's accuracy of 81 %.

This dataset is publicly available for download without any constraint. For a convenient bulk download, the dataset homepage provides a " *.txt " files comprising the Uniform Resource Locator (URL) of the tiles for each sub-dataset. The spatial location of an image can be directly determined from the file name within the URL. Each file name has a common pattern shown by the following expression: " Hansen_VERSION_LAT [NS] _LNG [WE] ". LAT (latitude) and LNG (longitude) refer to the top left corner coordinates of a raster image, whereas these coordinates are only given in natural numbers. The orientation of the image on the hemisphere is determined by the four cardinal directions N (north), S (south), W (west) and E (east). For this project, we require all three sub-datasets, namely: Treecover2000, lossyear, and gain. The data acquisition is automatized with an Python script by using the Standard Library (stdlib) modules urllib, re, and threading [van Rossum and Development 2018]. At first, the Python script downloads the provided " *.txt " files and creates a list data structure, where each URL is element of this list. After, it cycles through the list and extracts the corner coordinates from the file name by means of a Regular Expression (REGEX). These corner coordinates and cardinal directions are converted to valid latitude and longitude coordinates between [−90, 90] and [−180, 180], respectively. Now, an image is only downloaded if it is within the study extent between [−20, 30] latitude. The acquired image tiles in total 678 are shown in the top panel (green squares) in figure 2.1.

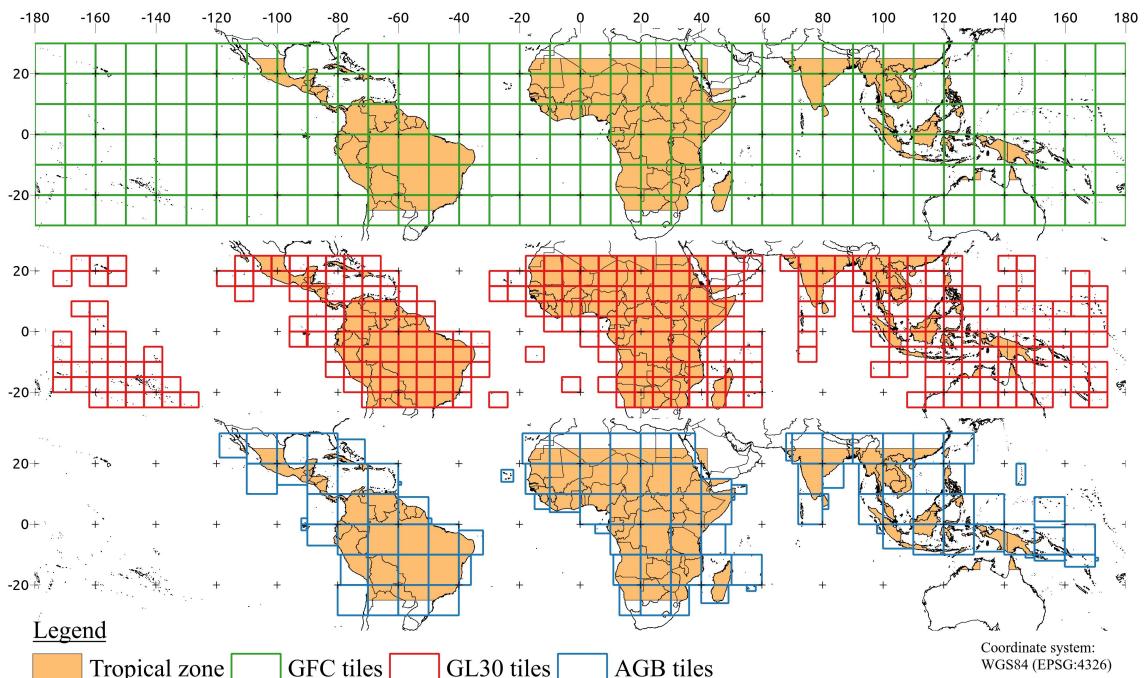


Figure 2.1. Map of downloaded dataset tiles: This map shows the acquired image tiles for this study. From top to bottom in green Global Forest Change (GFC) dataset tiles (Treecover2000, lossyear and gain), the land cover dataset GlobeLand30 (GL30) image tiles in red, and in blue the Aboveground Biomass (AGB) dataset tiles. The orange filled shapes highlight countries within the tropical zone.

2.1.1.2 GlobeLand30

GlobeLand30 (GL30) is the first global land cover dataset with a 30 meter per pixel spatial resolution that provides a comprehensive view on the distribution of 10 different land cover classes (table 2.1) over the entire globe [Chen et al. 2017]. Currently this dataset is available for two different time periods 2000 and 2010 [Chen et al. 2015]. The pixel values of this dataset are coded in unsigned 8 bit integers and as coordinates system it uses WGS84 in Universal Transverse Mercator (UTM) projection. GL30 can be downloaded as a GTiff raster mosaic where each image covers 6x5 degrees [Chen et al. 2014]. For detecting the land cover classes Chen et al. used a so called Pixel-Object-Knowledge (POK) oriented approach and satellite imagery from Landsat ETM+ [Chen et al. 2015]. Chen et al. divided the mapping process in different stages where each land cover type is detected separately and deleted subsequently from the source satellite image. The applied mapping order is: water bodies, wetland, snow and ice, cultivated land and forest, shrubland, grassland and bareland synchronous. To detect the pixels of a selected land cover type the following pixel level classifiers are used: Decision Tree (DT), Support Vector Machine (SVM) or Maximum Likelihood Classifier (MLC). After pixel detection the adjacent pixels are grouped as an aggregated land use object. These objects are subsequently validated by expert knowledge and the gained knowledge is used as a feedback loop to improve the automatized classification.

Chen et al. estimates an overall mapping accuracy of 80.33 % and 78.6 % for 2000 (only validated in Shaanxi, China) and 2010 (global), respectively [Chen et al. 2015]. Several research groups besides Chen et al. validated the mapping accuracy of GL30 at different regions and scales. Arsanjani et al. estimates an overall accuracy of 77.9 % for Iran and an accuracy >80 % for Germany [Arsanjani et al. 2016a,b]. Yang et al., Cao et al. and Jacobson et al. estimate an accuracy of 82.4 %, 80.1 % and 83.1 % for China, Nepal and East Africa, respectively [Yang et al. 2017; Cao et al. 2016; Jacobson et al. 2015]. Unfortunately, no study focused on validating the mapping accuracy for regions exclusively within the tropical zone.

Chen et al. donated the GL30 land cover mapping to the United Nations (UN) but it is not accessible for public download. The download is restricted to users who register on the dataset homepage but the registration process is not working properly. Fortunately the supervisor of this work had already an account otherwise it would be impossible to receive a copy of the dataset. A registered user must fill an order application to get access to the image tiles. The application form must contain the tile identifiers and the selected time period. Tile identifiers have the following common pattern: "[NS] ZONE_LAT_NAME" where zone refers to the UTM zone between [1, 60], N (north) or S (south) to the cardinal direction, and LAT (latitude) to the latitude coordinate of the top left corner. For a better usability the homepage provides an interface for selecting the required image tiles but the selection of multiple tiles

did not work. As well a vector file is provided which contains the dataset tile polygons with assigned identifiers. This file was used to select all required tiles within the tropical zone between approximately $[-23, 23]$ degrees (WGS84). Figure 2.1 presents the selected images in red at middle panel. The corresponding image identifiers are converted to an single line string and copied to the application form. After submitting the form the order will be checked and approved within two weeks. After one week we received a two weeks limited access to an password protected File Transfer Protocol (FTP) server where we downloaded 716 raster images. Due to the several restrictions this process of selecting and downloading could not be automatized with one pipeline. Only the selection and string conversion was automatized with a throw away script.

Table 2.1. Classification schema of the GlobeLand30 product: The code column is the assigned pixel value, type the corresponding land cover type and definition explains in broad terms which types of surfaces fall into the land cover type [Chen et al. 2017].

Code	Type	Definition
10	Cultivated land	used for agriculture, horticulture and gardens, including paddy fields, irrigated and dry farmland, vegetable and fruit gardens, etc.
20	Forest	covered by trees, vegetation covers over 30 %, including deciduous and coniferous forest, and sparse woodland with cover 10-30 %, etc.
30	Grassland	covered by natural grass with cover over 10 %, etc.
40	Shrubland	covered by shrubs with cover over 30 %, including deciduous and evergreen shrubs, and desert steppe with cover over 10 %, etc.
50	Wetland	covered by wetland plants and water bodies, including inland marsh, lake marsh, river floodplain wetland, forest/shrub wetland, peat bogs, mangrove and salt marsh, etc.
60	Water bodies	in land area, including river, lake, reservoir, fish pond, etc.
70	Tundra	covered by lichen, moss, hardy perennial herb and shrubs in the polar regions, including shrub-, herbaceous-, wet- and barren-tundra, etc.
80	Artificial surfaces	modified by anthropogenic influence, including all kinds of habitation, industrial and mining area, transportation facilities, and interior urban green zones and water bodies, etc.
90	Bareland	with vegetation cover lower 10 %, including desert, sandy fields, Gobi, bare rocks, saline and alkaline land, etc.
100	Snow and ice	covered by permanent snow, glacier and icecap

2.1.1.3 Intact Forest Landscapes

A Intact Forest Landscapes (IFL) is a mosaic of undisturbed forest patches and naturally treeless ecosystems without signs of human activity and large enough to maintain all native

biological diversity [Potapov et al. 2017]. Due to the fact that IFL comprises different intact natural landscape patterns like primary forests, non-forest ecosystems, temporary treeless areas after a natural disturbance, and water bodies the term is not congruent to the term primary forest defined by the Food and Agriculture Organization of the United Nations (FAO) [FAO 2012]. But as mentioned IFLs includes large patches of primary forests with a minimum extent of 500 Km² therefore primary forests can be extracted from the layer. Still there are smaller fragments of primary forest outside of the IFLs. In regards of the extent an IFL has a minimum size of 500 Km², a minimum width of 10 Km, and a minimum corridor/appendage width of 2 Km. Further an IFL should not contain any of the following: ecosystem alteration, fragmentation by infrastructure and disturbance, and areas altered or managed through agriculture, logging, and mining. For mapping and detecting IFLs Potapov et al. used Landsat imagery and several auxiliary data sources like GFC, and national transportation maps. The dataset can be downloaded as a Shapefile (SHP) file with the coordinate reference system WGS84. Each polygon in the SHP represents an IFL patch at a certain location on our planet at the time period 2000.

Data acquisition is pretty straight forward the IFL dataset public accessible for download. As mentioned it is an SHP so you must only download a single compressed archive. The download is automatized with an Python script by using the stdlib modules urllib and threading [van Rossum and Development 2018].

2.1.1.4 Aboveground Woody Biomass

The Aboveground live woddy Biomass density (AGB) raster dataset is prepared by Global Forest Watch (GFW) by an adapted approach of Baccini et al. [Baccini et al. 2012, 2015, 2017]. For the year 2000, this dataset estimates the aboveground biomass density per pixel in Mg C ha⁻¹ (mega gram carbon per hectare), and the confidence per pixel at a spatial resolution of approximately 1 arc-second (approximately 900 Km² or 30x30 meter). The dataset covering the global tropical zone as an mosaic of GTiff raster images where each tile of the mosaic has the Coordinate Reference System (CRS) WGS84 and is coded in float. For deriving biomass density GFW used canopy metrics from Geoscience Laser Altimeter System (GLAS) Light Detection and Ranging (LIDAR) footprints and several regional and forest specific allometric equations. The resulting GLAS AGB estimates are used as labels to train regional specific Random Forest (RF) models based on Landsat 7 ETM+ top-of-atmosphere reflectance, tree canopy density of GFC, elevation data, and climate data as predictor variables. After these models are subsequently applied to the entire study extent to predict the biomass content for each pixel. Additional a uncertainty layer is prepared accounting for the errors from allometric equations, the LIDAR based model, and the random forest model.

The AGB raster mosaic is public available on the homepage of GFW. As mentioned, the dataset covers only the tropical zone, therefore we acquire the entire mosaic. The GFW homepage provides an Geographic JavaScript Object Notation (GeoJSON) Application Programming Interface (API) to receive the actual URL of each raster image. If a request is sent to this API the server response with a GeoJSON feature collection. The collection contains as attributes the URLs of the biomass images, the URL of the uncertainty layers, and the rectangular bounds of each image. The data acquisition is automatized by means of Python and the stdlib modules urllib, threading, and the open source library geopandas [van Rossum and Development 2018; McKinney 2010]. At first the GeoJSON is downloaded via an API call and eventually stored on disk. Next we iterate the features of the GeoJSON collection and extract the URLs (biomass and uncertainty) of each tile. These URLs are downloaded and subsequently stored on disk. During the downloads of the uncertainty layers the GFW server answered repeatedly with a 404 (Not found). Therefore the uncertainty layers are not available. In total we downloaded 105 different image tiles, their extent and spatial location is shown in blue at the bottom panel of figure 2.1.

2.1.1.5 Global Soil Organic Carbon

The Global Soil Organic Carbon map (GSOCmap) is a joint project between Global Soil Partnership (GSP) and Intergovernmental Technical Panel on Soils (ITPS) to produce a global Soil Organic Carbon (SOC) content map by a country driven approach. In the year 2018, the first iteration of this map in version 1.0 was released, and shortly followed by 1.1 (new country submission by Rwanda) and 1.2 (new country submissions by Chile and Colombia). As the short release cycle suggests the mapping project is intended as a long-lived dataset which will improve over time and by new country submissions. Till now 67 (approximately 63 % of the global land mass) different countries submitted their country based SOC estimates. To foster the national SOC mappings the International Soil Reference and Information Center (ISRIC) provides several covariate datasets like national Digital Elevation Map (DEM) maps, annual spectral remote sensing data or national soil type grids. Additionally the contributors can join a mapping training and use the GSOCmap cookbook as guidance for their mapping efforts. As an exchange, each country shares its national GSOCmap by compliance of several criteria e.g. reporting of the Meta-data of the SOC sampling (sample timeline, sample depth, bulk density etc.), uncertainty assessment, and the applied methods for estimating and interpolation of the SOC content. For interpolating the guide organizations suggest the following approaches: simple geo-matching, class-matching, Multiple Linear Regression (MLR), RF or SVM. The national maps are aggregated to the final GSOCmap with a target resolution of 30 arc-seconds (approximately 1 Km²) in the CRS WGS84. The dataset is one single raster image as GTiff coded in float covering the entire

globe where each pixel value is the SOC content in Mg C ha⁻¹ at a soil depth of 0-30 cm [FAO and ITPS 2018].

The product is validated by comparing the pixel level estimates with soil sampling data from various soil databases (WoSIS, HWSD, etc.). In total 312122 samples were divided into three sub-levels (<150 Mg C ha⁻¹, >150 Mg C ha⁻¹, and all samples) and subsequently computed the Mean Error (ME). The ME of the entire sample space and <150 Mg C ha⁻¹ suggests that the mean Soil Organic Carbon Content (SOCC) estimate is an overestimate of 1.6 and 4.5 Mg C ha⁻¹ respectively. All samples with a SOCC content >150 Mg C ha⁻¹ show an underestimate by approximately 165 Mg C ha⁻¹ in the mean. Additionally, an uncertainty assessment was prepared to estimate a Standard Deviation (SD) between ± 0-16 t ha⁻¹ for the tropical zone. Unfortunately, this assessment is pretty rough and till now not available as a product. The GSOCmap in comparison with other global SOC products has the lowest Root Mean Square Error (RMSE). In summary, the prepared validations show evidence that the GSOCmap is a conservative data product with a tendency to underestimate the SOCC.

The dataset is publicly available at the homepage of the FAO. As mentioned it consists of one raster image, therefore we download it by means of a Python script without any additional steps.

2.1.1.6 Auxiliary

As auxiliary data for country boundaries we downloaded with Python the Global Administrative Areas Map (GADM) layers as SHP files [Hijmans et al. 2018; van Rossum and Development 2018].

2.1.2 Empirical data

2.1.2.1 Soil Organic Carbon

Don et al. performed the first study of tropical SOC change for soil depth between 0 and 30 cm. For the study a global meta-analysis is applied by using 358 (153 published and peer-reviewed) different studies to estimate SOC change for 12 major Land-use Change (LUC) types. The base date is derived from 39 different tropical countries covering all continents. Unfortunately Africa and East-Asia are under-sampled whereas South-America have the best data coverage. The meta-analysis is restricted to mineral soils therefore all wet soil types are excluded from the analysis Don et al.. The 12 Land Cover (LC) transitions encompass the following LC types: primary forest, secondary forest, grassland, cropland, and perennial

crops. Primary forest are defined as natural vegetation without human impacts which includes natural grassland and shrubland. Secondary forest are managed forests and regrown forests after partial destruction of the old stand. Grassland comprises pastures for livestock but excludes natural grasslands. Cropland comprises annual crops like maize or beans and perennial crops could be coffee or sugar cane. For our study we used only the SOC change estimates for these LUC types which corresponds to the GL30 and IFL classification schema. The actual values are shown in table 2.2.

Table 2.2. Relative soil organic carbon change for certain land-use change types: The Land-use change columns from and to define the LUC type with the corresponding relative Soil Organic Carbon (SOC) change and the Standard Error of the Mean (SEM) [Don et al. 2010].

LUC type		Relative SOC change	
From	To	[%]	SEM
Primary forest	Grassland	-12.1	±2.3
Primary forest	Cropland	-25.2	±3.3
Primary forest	Secondary forest	-8.6	±2.0
Secondary forest	Grassland	-6.4	±2.5
Secondary forest	Cropland	-21.3	±4.1

2.1.2.2 Ecosystem Service Values

Some text to ensure document flow.

Table 2.3. Selection of Ecosystem Service Values (ESV) per biome used in this study: ESV per biome connected with the corresponding GlobeLand30 land-cover class, and its monetary value in Int.\$ ha⁻¹. Dg refers to data from de Groot et al., Co from Costanza et al., and Wb from Siikamaki et al.[de Groot et al. 2012; Costanza et al. 2014; Siikamaki et al. 2015]

Biome	Code	Type	Dg	Co	Wb
Cropland	10	Cropland	-	5,567	-
Forest tropical	20	Forest	5,264	5,382	1,312
Forest tropical	25	Regrowth	5,264	5,382	1,312
Grass/Rangelands	30	Grassland	2,871	4166	-
Wetlands	50	Wetland	25,682	140,174	-
Lakes/Rivers	60	Water bodies	4,267	12,512	-
Urban	80	Artificial	-	6,661	-

2.2 Methods

2.2.1 Preprocessing

Before we apply further analysis, we have to generalize the used datasets. As introduced in the data section do we use datasets which differ largely in their metadata properties, for example, single-tiled or multi-tiled images, used CRS, spatial resolution, and file type. Therefore, our goal should be to develop a process which creates an image stack of equal meta-data for each location in our study extent. In further descriptions, we will refer to this stack as Aligned Image Stack Mosaic (AISM). As target CRS for our AISM we chose WGS84 and as target extent for the mosaic, we use the bounding box of the GL30-2010 tiles. The following paragraph explains how we developed the alignment algorithm by means of Python and the additional open source libraries rasterio, geopandas, and shapely [van Rossum and Development 2018; McKinney 2010].

The first exercise of the preprocessing algorithm is to detect all tiles covering the extent of our template tiles. At first, we create for each multi-tiled dataset a polygon mask as SHP. This mask contains the spatial extent of each tile within a dataset and as attribute the corresponding file identifier. If the dataset tiles are not in WGS84 the extracted bounds are subsequently reprojected to this CRS. During the masking process, we recognized that the raster mosaic bounds of both GL30 datasets (2000 and 2010) generate re-projection errors. These errors showed up as polygons spanning the entire globe but one tile can only fill its UTM zone extent. A further analysis revealed that all tiles located in UTM zone 1 and 60 overflowing the maximum and minimum longitude coordinates of this zones. As solution we excluded all tiles within UTM zone 1 and 60 from further processing, namely: n01_00, s01_00, s01_10, s01_15, s01_20, s60_00, s60_05, s60_10, s60_15, and n53_00. The described steps can be found as well in figure 2.2. Now, as the figure suggests we determine the intersection between these mask layers and group the intersecting tiles by our template tile. Next, we create for the template tile a re-projection profile (warp profile) and apply it subsequently to all intersecting tiles based on the following rules: if from one dataset more than one tile intersects merge them followed by re-projection or if only one tile intersects just re-project it. As introduced the GSOCmap consist only of one single tile with a spatial resolution of approximately 1 Km², so it must only re-project and re-sampled by nearest-neighbor approach. We select from the IFL layer all polygons within our template warp profile and convert them to a raster layer where intact forest patches are coded by a one in 8-bit unsigned integer. Last step of the alignment process is the rounding of the AISM bounds to full integer degrees and a subsequent clipping of each tile to this rounded bounds. The entire work flow is pictured in figure 2.2 and results in a AISM shown in figure 2.3. Finally,

we create a polygon mask of our AISIM and store for each polygon the corresponding dataset tiles. This mask is used as a file index for the next algorithms.

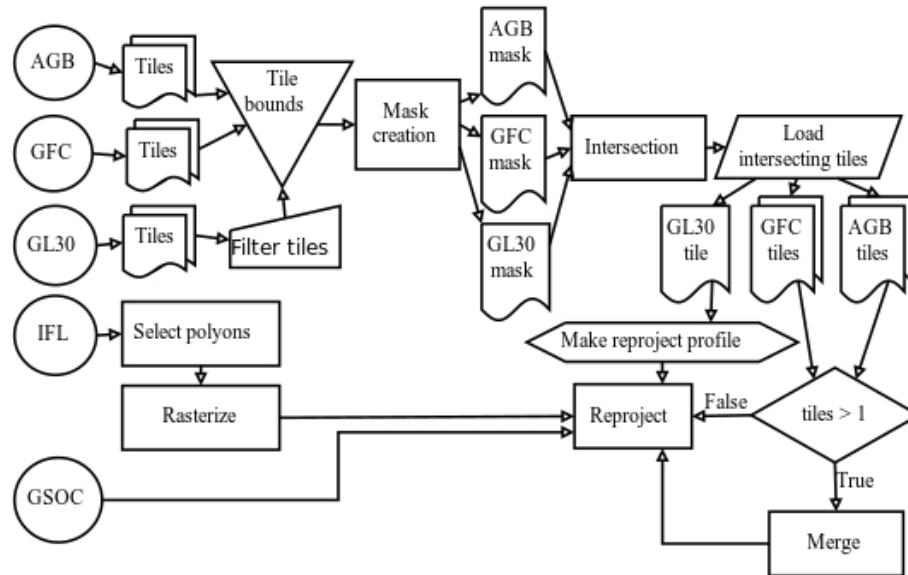


Figure 2.2. Flowchart of tile alignment process: For the multi-tiled datasets (multi-document symbols) a mask is created by extracting the tile bounds. Next, the intersection between these masks is determined and the corresponding tiles are loaded from disk. GL30 tiles are used as template by creating re-project profile and subsequently applying it to intersecting tiles. From the IFL layer only polygons within the re-project are selected and subsequently converted to a raster layer.

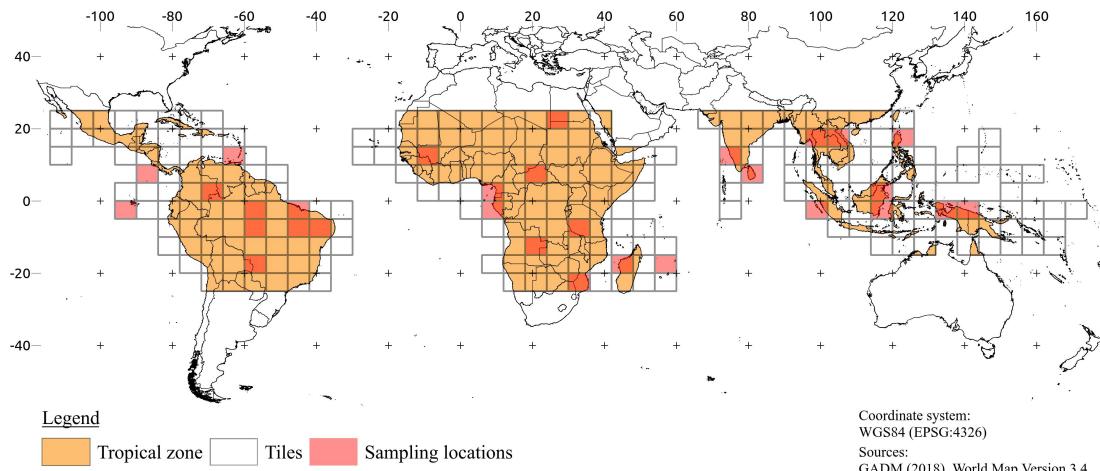


Figure 2.3. Map of aligned data tiles and sampling locations: The map shows the location of the aligned multi-image stack tiles as black-framed square sized polygons, the sampling locations for accuracy assessment in red, and countries within the tropical zone in orange.

2.2.2 Deforestation

2.2.2.1 Forest definition

Goal: We aggregate two layers. Both have slightly different definitions for tree cover. To successful aggregate the layers we must harmonize the tree cover situation between the gfc tree cover and the gl30 2000 treecover. Because we want to map changes both have to agree on tree cover.

Similarity: To determine the similarity between gl30 and gfd tree cover we used jaccard index. Jaccard Index (coefficient of community) is a simple measure of similarity between two binary paired populations or the measure of the degree of spatial overlap between two images. First applied by Jaccard in 1912 to compare distributions of alpine flora. Properties of Jaccard if a is zero jaccard is zero if b and c are zero jaccard is one. value range between 0 and 1 and relationship between $s(ij)$ and a near linear. Process we computed subsequently for each image tile the jaccard index. First step is to filter gl30 map for tree cover. Next we compute the jaccard index The process is implement as binary operations ($a = x \text{ or } y$, $abc = x \text{ or } y$, $b = x \text{ xor } a$, $c = y \text{ xor } a$, $d = \text{not } x \text{ and not } y$) by means of numpy and python.

Hypothesis testing:

Treecover agreement: Jaccard index first applied for a study on plant distribution similarities at different locations in alps and developend by jaccard [Jaccard 1912; Shi 1993; Sampat et al. 2009] jaccard index, wilcoxon signed-rank test, gl30 2000, gfc treecover 2000, different canopy densities $JI_0[0, 100]$, $JI_1(10, 100]$, $JI_2(20, 100]$, $JI_3(30, 100]$

2.2.2.2 Proximate deforestation driver

Based on our forest definition developed in the previous section we want to classify all the tropical deforestation within a canopy density of $(10, 100]$ percent between 2001 till 2010. Additionally we must consider the mean miss-classification rate of 52 % by previous findings of Seydewitz [Seydewitz 2017]. Therefore we have to develop a feasible method to resolve this issue.

For classifying the proximate drivers of deforestation we select the following raster images from our AISM: GFC treecover2000, GFC lossyear, GFC gain, and the GL30 LC classification from 2010. Now, we apply to each raster image stack the following described operations. From the treecover2000 layer we select all pixels where the canopy density is within $(10, 100]$ percent and set them to one (true). The same is procedure is applied on the lossyear layer by setting all forest loss pixels within the time period 2001 till 2010 to one (true). After both layers are combined with a logical AND to select our target deforestation

pixels. Finally, we build the Hadamard product (element-wise multiplication) of the target deforestation layer and the GL30 LC layer to classify the deforestation pixels. For classifying forest regrowth we filtered the GFC gain layer by forest gains within our target temporal resolution and losses within our target canopy density. After, the filtered gains are aggregated with our classified deforestation by using the Hadamard product of both layers. Figure 2.4 shows the classification process as a flowchart. The classification process is implemented as a Python function which requires as parameter the previously named raster layers. Additionally the target canopy density and time period is freely selectable. The described filtering and aggregation steps are implemented as binary operations for fast processing by means of numpy.

Clustering of all patches still classified as forest (value 20), apply square sized buffer around cluster center, count occurrences of other classes within the buffer, apply to cluster most frequent class, exclude from counting 20, 0, and 255, buffer size 500x500 m 25 ha.

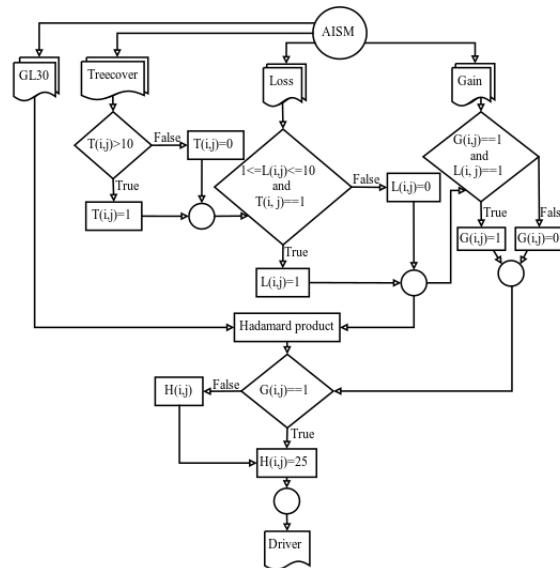


Figure 2.4. Title:

2.2.2.3 Accuracy assessment

For examining the accuracy of our proximate deforestation driver prediction we used a confusion matrix (also known as two-way frequency tables, error matrix or contingency tables). These matrices are commonly used for an accuracy assessment of land cover classifications and enables the computation of marginal and conditional distributions [Congalton 1991; Foody 2002]. Table 2.4 shows a general model of a confusion matrix. Foundation for an accuracy assessment by means of a confusion matrix is a collection of ground-truth samples

which can be compared with the class predictions for these samples produced by a classification algorithm. For the preparation of our accuracy assessment, we have to extract a collection of pixel samples with a deforestation occurrence from our proximate driver maps (further also called predictions). Next, we compose a set of ground-truth for these predictions (further also called references).

Table 2.4. A general model of a confusion matrix: X_1, \dots, X_n denote classification categories of two independent raters. $x_{n,n}$ are the actual samples sorted into the categories where the values in the diagonal show the agreement between both raters. The remaining cell values account for the disagreement between the two raters. Σ column and row show the marginal distribution and N is the total number of samples.

		Reference				
		Cls	X_1	\dots	X_n	Σ
Predict	X_1	$x_{1,1}$	\dots	$x_{1,n}$	$x_{1\cdot} = \sum_{i=1}^n x_{1,i}$	
	\vdots	\vdots	\ddots	\vdots	\vdots	
	X_n	$x_{n,1}$	\dots	$x_{n,n}$	$x_{n\cdot} = \sum_{i=1}^n x_{n,i}$	
	Σ	$x_{\cdot,1} = \sum_{i=1}^n x_{i,1}$	\dots	$x_{\cdot,n} = \sum_{i=1}^n x_{i,n}$	$\Sigma\Sigma = N$	

To create our collection of ground-truth data we draw randomly 10 image tiles from all three continental regions (Latin America, Africa, Asia/Oceania). From each tile, we sampled by random 200 pixels which total to 6000 samples over the entire study region. The sampling is realized with our own raster sampling algorithm build in Python by means of the open source libraries numpy and rasterio. As mentioned in the previous section do we superimpose two datasets and only a certain amount of pixels per tile is classified as proximate driver. Therefore, the sampling algorithm should only draw samples from occupied/classified pixels without replacement. The algorithm expects as parameters a raster image, the total number of samples to draw, a list of pixel values which should be interpreted as occupied cells, the affine transformation matrix of the raster image, and a seed for the random number generator. If occupied cells are set the algorithm will create a binary mask where each occupied cell is set to one relative to the input raster image. Otherwise it sets all pixel values greater or less than zero to one. After, the row and column coordinates of each one are extracted from the mask and converted to a flat list of coordinate tuples. Next, it draws the predefined number of samples from the list by a random order and uses the image coordinates to get the pixel value from the raster image. If a affine transformation matrix is provided the image coordinates are converted to real world coordinates. The seed argument ensures that on every algorithm rerun the samples are drawn. For our sampling we set the parameters to the following values: samples 200, occupied pixels GL30 class values and 25 for regrowth, affine matrix of the corresponding raster image, and the seed is 42. The per tile samples are stored as an Comma

Separated Values (CSV) file.

For the collection of ground-truth data we used visual interpretation of Google Earth satellite and aerial imagery. We developed a small JavaScript web application to access the imagery via the Google Maps API. The application expects as input a CSV file with the sampling coordinates. After upload of a sample file the user can cycle through the entries and the map jumps automatically to the coordinates of the sample. Now a reference label can be assigned to the coordinates by visual interpretation of the imagery. We subsequently assigned to all 6000 samples a reference label and downloaded the results as CSV.

Finally, we developed a Python class to compute the confusion matrix. The constructor of the class requires a list of reference and prediction labels. With the provided arguments it creates the confusion matrix. Further, it computes the following marginal and conditional distributions: overall accuracy $OvAc$ by dividing the sum of classification agreements through the sample total N (equation 2.1), the producer accuracy $PAc.n$ by dividing the category agreement through the column category total (equation 2.2), the error of commission $Com.n$ (Type II error) by dividing the category disagreement through the column category total (equation 2.3), the user accuracy $UAc.n$ by dividing the category agreement through the row category total (equation 2.4), the error of omission $Om.n$ (Type I error) by dividing the category disagreement through the row category total (equation 2.5), and the Cohens Kappa by substituting equation 2.6 and 2.1 into equation 2.7.

$$p_0 = OvAc = \frac{\sum_{i=1}^n x_{i,i}}{N} \quad (2.1)$$

$$PAc.n = \frac{x_{i,i}}{x_{.n}} \quad (2.2)$$

$$Com.n = \frac{FN_i}{x_{.n}} \quad (2.3)$$

$$UAc.n = \frac{x_{i,i}}{x_{n.}} \quad (2.4)$$

$$Om.n = \frac{FP_i}{x_{n.}} \quad (2.5)$$

$$p_c = \frac{1}{N^2} \sum_{i=1}^n x_{i.} \cdot x_{.i} \quad (2.6)$$

$$Kappa = \frac{p_0 - p_c}{1 - p_c} \quad (2.7)$$

2.2.3 Emissions

2.2.3.1 Above ground biomass

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2.2.3.2 Soil organic carbon change

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sociis natoque penatibus et magnis dis parturient montes, nascetur ridiculus mus. Aliquam tincidunt urna. Nulla ullamcorper vestibulum turpis. Pellentesque cursus luctus mauris.

2.2.4 Ecosystem service values

2.2.4.1 Ecosystem service value loss

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2.2.4.2 Ecosystem service value gain

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2.2.5 Binning analysis

The previous sections were focused on the generation of large scale spatial data. Now, a feasible method must be developed for analyzing, aggregating, interpreting, and visualizing our results. For the development of a proper approach we have to generalize the problem domain. At first we are confronted with large N (many samples) which results in a high dimensionality and complexity of relationships among this variables [Carr 1990]. From a visual/analytical perspective georeferenced raster maps can be interpreted as a multivariate scatter plot of large datasets where longitude and latitude represent the x and y coordinate of an data point and the pixel values (in this case nominal scaled) representing the third dimension as an group coloring. Therefore we have a large multidimensional dataset combined with a scatter plot visualization which leads commonly to over plotting issues and hidden point densities [Carr et al. 1987]. Due to the spatial nature of your data we are also confronted with not equal distributed data some regions show high data densities and other regions have sparse to no data. Also a severe problem domain is the frame size of our representation. Goal is to present data on a continental level which intensifies visual problems. Each pixel has a resolution of approximately 30x30m, the continental representation of americas spanning approximately 1200000x120000km². Therefore small scale isolated changes are hidden and only large scale changes are visual detectable. Which results in hidden details and not perceivable patterns of change.

Goal should be to develop an process who solve this issues and generates satisfying output for our multivariate data. In case of raster data a re-sampling to coarser resolution could solve over plotting and resolution issues as well normalize the unequal distributed data. But the nature of re-sampling (for nominal data a nearest neighbor or majority wins [Reference](#)) would negate important spatial patterns as well frequency distributions. Another well known approach is to use binning of the spatial explicit data with a certain kind of regular polygon that is tessellating the plane [Carr et al. 1992]. Polygon tessellations provide numerous opportunities for presenting multivariate statistical summaries. The scaling of the polygon could be used to represent pixel densities within the polygon area, a polygon filling color gradient is applicable to show nominal or ordinal scaled data. Also it is imaginable to use the polygon interior for a pie chart. To use regular tessellation it is important to mention there are only three types of regular polygons tessellate the plane: squares, equilateral triangles and hexagons [Carr et al. 1992]. Square tessellation is the most common approach used for binning in spatial visualization. A raster image is a square tessellation. In a square mosaic each polygon shares 4 edge neighbors and 4 vertex neighbors [more explanation error distance disadvantages etc Hexagons properties, advantages disadvantages of both tessellations](#). Final goal is to show your analysis results of spatial explicit raster data in hexagonal binned form. For bivariate maps we choose a visual representation with scaled hexagons and colorization.

For multivariate details we choose a pie chart alike visualization. We split the hexagons horizontal in regards of the presented ratio. The ratios should be ordered descending so that the greatest ratio is south oriented. It is following a general description how we created the hexagon grids and how we tackled the polygon split problem.

To be flexible at hexagon construction we accept 4 different parameters as construction arguments: D long diagonal (Diameter of the circumscribing circle), d short diagonal (diameter of the inscribed circle), A area the hexagon should span and or e the edge length. One selected parameter of these is used to compute R the radius of the circumscribing circle with respect to input parameter as shown in equation 2.8. R is used to calculate the midpoint $\langle c_x, c_y \rangle$ of the hexagon located in the first quadrant of the cartesian coordinate system Equation 2.9 and 2.10. Equation 2.11 shows the computation of the hexagon anti-clockwise vertex matrix. Whereas the two leftmost vertices (first and last row of the matrix \mathbf{H}) are located at koordinatenursprung, will sagen auf deutsch korridanten at $x=0$ und y =value of matrix. In summary equation 2.8 to 2.11 show the creation of an hexagon at the leftmost corner of first quadrant (Figure 2.5). The orientation is important for the subsequent mosaic creation.

$$R = \frac{\sqrt{2A}}{\sqrt[4]{27}} = \frac{D}{2} = \frac{d}{\sqrt{3}} = e \quad (2.8)$$

$$c_x = \frac{R\sqrt{3}}{2} \quad (2.9)$$

$$c_y = R \quad (2.10)$$

$$\mathbf{H} = \begin{bmatrix} 0 & c_x & 2c_x & 2c_x & c_x & 0 \\ R\sin\left(\frac{7\pi}{6}\right) + c_y & 0 & R\sin\left(\frac{11\pi}{6}\right) + c_y & R\sin\left(\frac{\pi}{6}\right) + c_y & 2R & R\sin\left(\frac{5\pi}{6}\right) + c_y \\ 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix} \quad (2.11)$$

A polygon tessellation needs several polygons to create a grid in case of the creation of one hexagon with the presented algorithm needs approximately **benchmark** but the creation of **several N hexagons** needs approximately **benchmark**. Therefore it is much simpler to create only one hexagon with the presented algorithm and to create the grid polygons by copying the coordinates of the source polygon and translating them to their target position with a affine transformation matrix shown in equation 2.12. To create the grid we get the rectangular bounds of the area to tessellate as a matrix $\mathbf{B} \in R^{2 \times 2}$ (equation 2.13), where the first column of the matrix contains the lower left corner and the second column the upper right corner of the image. Each subsequent translation in regards of x_{off} is $x_1 + d$ for even rows and bla bla for odd rows. Y_{off} is computed by bla bla see figure 2.5.

$$\mathbf{T} = \begin{bmatrix} 1 & 0 & x_{off} \\ 0 & 1 & y_{off} \\ 0 & 0 & 1 \end{bmatrix} \circ \mathbf{H} \quad (2.12)$$

$$\mathbf{B} = \begin{bmatrix} x_1 & x_2 \\ y_1 & y_2 \end{bmatrix} \quad (2.13)$$

Binning of raster data is easy we just have a point in polygon problem each points/pixels

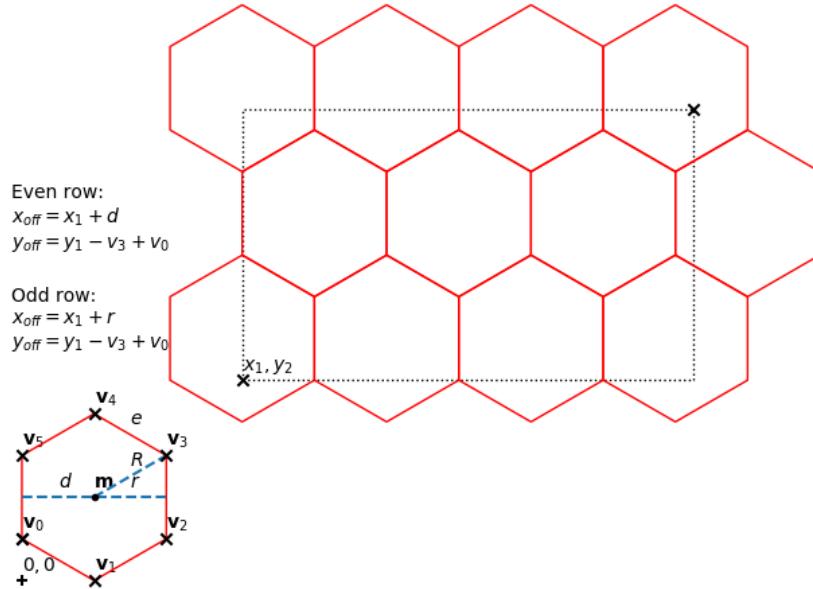


Figure 2.5. Hexagon tessellation: Located at the left bottom corner in red a hexagon defined by its geometric properties the 6 vertex vectors $\{\vec{v}_0, \dots, \vec{v}_5\}$ (black crosses), with center vector \vec{m} , edge length e , R radius of the circumscribing circle, r radius of the inscribed circle and d the length of the short diagonal. Top right black dotted box are the bounds of an area which is tessellated by a hexagon grid in red. Each grid cell is translated from the origin hexagon at its position by computing the x_{off} and y_{off} offset with the presented equations at the left-hand side of the grid.

falling in hexagon are counted and aggregated through a function. In case of drivers of deforestation we count all driver classes per hexagon and compute ratios next we compute the sha **describe for each map how you build it** As mentioned before for the visualization of the drivers of deforestation map we want to segment the hexagons with horizontal lines and each segment should represent the share of the direct deforestation driver within the tessellated area. To compute the split line for a certain hexagon we need the hexagon R computeable from the area of the hexagon equation 2.14 and the rectangular bounds of the hexagon. We compute the relative share of an deforestation driver per hexagon this relative share can be used to compute the y-axis coordinate of an split line equation 2.15. A regular hexagon can not only be presented in it vertex form as shown above. We can also use functions to define the hexagon shape. A hexagon consist of 2 picewise functions where each function consist of 3 linear functions restricted to an intervall. If we invert these functions we can use these functions to compute the x-coordinate of the split line with the previous computed y-coordinate Equations 2.16 and 2.17. As a results we receive the solution matrix L which represents the horizontal line segment splitting the hexagon at the point where we want (driver ratio share) equation 2.18. The solution matrix can be plugged in to a polygon

split function which separates the hexagon polygon in a upper and lower part to do so we iterate over the hexagon vertices and decide if they are above or under the split line and append to a lower upper polygon. These list are our results [explain better split function](#).

$$R = \frac{\sqrt{2A}}{\sqrt[4]{27}} \quad (2.14)$$

$$y = \frac{P(y_2 - y_1)}{100} + y_1 \quad (2.15)$$

$$f^{-1}(y) = \begin{cases} -\frac{y-y_1}{\tan(\frac{\pi}{6})} + \frac{x_1+x_2}{2} & \text{if } y_1 \leq y < y_1 + R \sin(\frac{5\pi}{6}) \\ x_1 & \text{if } y_1 + R \sin(\frac{5\pi}{6}) \leq y < R(\sin(\frac{5\pi}{6}) + 1) \\ \frac{y-y_2}{\tan(\frac{\pi}{6})} + \frac{x_1+x_2}{2} & \text{if } R(\sin(\frac{5\pi}{6}) + 1) \leq y \leq y_2 \end{cases} \quad (2.16)$$

$$g^{-1}(y) = \begin{cases} \frac{y-y_1}{\tan(\frac{\pi}{6})} + \frac{x_1+x_2}{2} & \text{if } y_1 \leq y < y_1 + R \sin(\frac{5\pi}{6}) \\ x_2 & \text{if } y_1 + R \sin(\frac{5\pi}{6}) \leq y < R(\sin(\frac{5\pi}{6}) + 1) \\ -\frac{y-y_2}{\tan(\frac{\pi}{6})} + \frac{x_1+x_2}{2} & \text{if } R(\sin(\frac{5\pi}{6}) + 1) \leq y \leq y_2 \end{cases} \quad (2.17)$$

$$\mathbf{L} = \begin{bmatrix} f^{-1}(y) & g^{-1}(y) \\ y & y \end{bmatrix} \quad (2.18)$$

3 Results

3.1 Forest definition and accuracy assessment

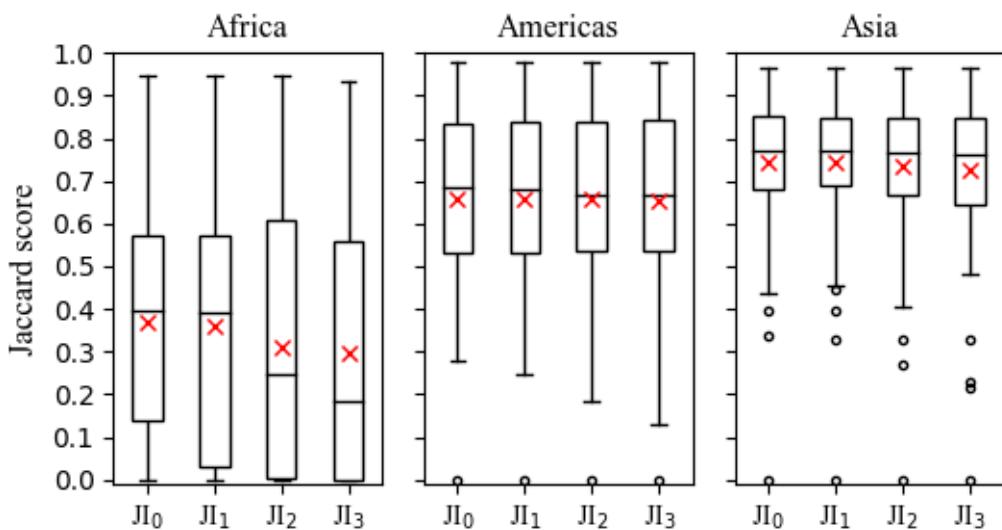


Figure 3.1. Similarity distribution of GFC tree cover vs. GL30 tree cover at 2000: Americas from JC₀ till JC₃: $\bar{x} \approx .71, .71, .70, .69$; $\tilde{x} \approx .77, .77, .76, .75$; $Q_1 \approx .64, .65, .64, .63$; $Q_3 \approx .84, .84, .84, .83$; Asia from JC₀ till JC₃: $\bar{x} \approx .71, .71, .70, .69$; $\tilde{x} \approx .77, .77, .76, .75$; $Q_1 \approx .65, .66, .64, .63$; $Q_3 \approx .84, .84, .84, .83$; Africa from JC₀ till JC₃: $\bar{x} \approx .39, .39, .34, .32$; $\tilde{x} \approx .40, .40, .31, .29$; $Q_1 \approx .14, .10, .00, .00$; $Q_3 \approx .60, .61, .63, .56$

Table 3.1. Two-sided Wilcoxon signed-rank test:

$H_0: \tilde{x}_1 = \tilde{x}_2$													
Cls	Americas				Asia				Africa				
	Jl ₀	Jl ₁	Jl ₂	Jl ₃	Jl ₀	Jl ₁	Jl ₂	Jl ₃	Jl ₀	Jl ₁	Jl ₂	Jl ₃	
Jl ₀	1.					1.				1.			
Jl ₁	.00	1.				.72	1.			.22	1.		
Jl ₂	.06	.36	1.			.00	.00	1.		.03	.03	1.	
Jl ₃	.16	.50	.60	1.		.00	.00	.00	1.	.00	.00	.00	1.

Table 3.2. One-sided Wilcoxon signed-rank test:

$H_0: \tilde{x}_1 \leq \tilde{x}_2$													
		Americas				Asia				Africa			
	Cl	Jl ₀	Jl ₁	Jl ₂	Jl ₃	Jl ₀	Jl ₁	Jl ₂	Jl ₃	Jl ₀	Jl ₁	Jl ₂	Jl ₃
\tilde{x}_2	Jl ₀	1.	.00	.03	.08	1.	.64	1.	1.	1.	.11	.98	1.
\wedge	Jl ₁	1.	1.	.18	.25	.36	1.	1.	1.	.89	1.	.99	1.
\tilde{x}_1	Jl ₂	.97	.82	1.	.30	.00	.00	1.	1.	.02	.01	1.	1.
$H_0: \tilde{x}_1 \geq \tilde{x}_2$	Jl ₃	.92	.75	.70	1.	.00	.00	.00	1.	.00	.00	.00	1.

Table 3.3. Confusion matrix for accuracy assessment

		Reference												
		Cl	10	20	25	30	40	50	60	80	90	Tot	UAc	Om
Prediction	10	732	38	62	15	16	2	3	5	0	873	.84	.16	
	20	42	751	57	189	31	12	0	17	4	1103	.68	.32	
	25	29	202	1155	173	22	10	5	11	4	1611	.72	.28	
	30	36	187	32	1466	73	21	0	17	0	1832	.80	.20	
	40	14	21	4	41	352	1	1	2	1	437	.81	.19	
	50	0	5	3	10	4	50	0	1	0	73	.68	.32	
	60	2	1	0	3	0	2	18	2	0	28	.64	.36	
	80	3	4	0	1	1	1	0	50	0	60	.83	.17	
	90	0	0	0	1	0	0	0	3	5	9	.56	.44	
	Tot	858	1209	1313	1899	499	99	27	108	14	6026			
	PAC	.85	.62	.88	.77	.71	.51	.67	.46	.36			OvAc	
	Com	.15	.38	.12	.23	.29	.49	.33	.54	.64			.75	

3.2 Deforestation drivers

3.2.1 Global

3.2.2 Americas

3.2.3 Asia

3.2.4 Africa

Table 3.4. Absolute in km²

Class	Code	Type	Americas	Asia	Africa
Agriculture	10	Cropland	rel. abs.	24.37 95908	18.37 38719
					44368
Forestry/Plantations	30	Grassland	rel. abs.	46.19 181781	8.41 17726
					50.46 89516
Urban/Mining	25	Regrowth	rel. abs.	14.40 56671	70.27 148111
					18.61 33014
Natural	40	Shrubland	rel. abs.	12.69 49941	1.11 2340
					3.77 6688
Forest loss	80	Artificial	rel. abs.	0.41 1614	0.46 970
					0.71 1260
Forest cover	90	Bareland	rel. abs.	0.10 394	0.03 63
					0.09 160
Forest loss	50	Wetland	rel. abs.	1.50 5903	0.97 2045
					1.23 2182
Forest cover	60	Water	rel. abs.	0.32 1259	0.38 801
					0.13 231
Forest loss			rel. abs.	3.87 393550	4.68 210774
Forest cover			abs.	10223187	177400
			abs.	4457940	10496591

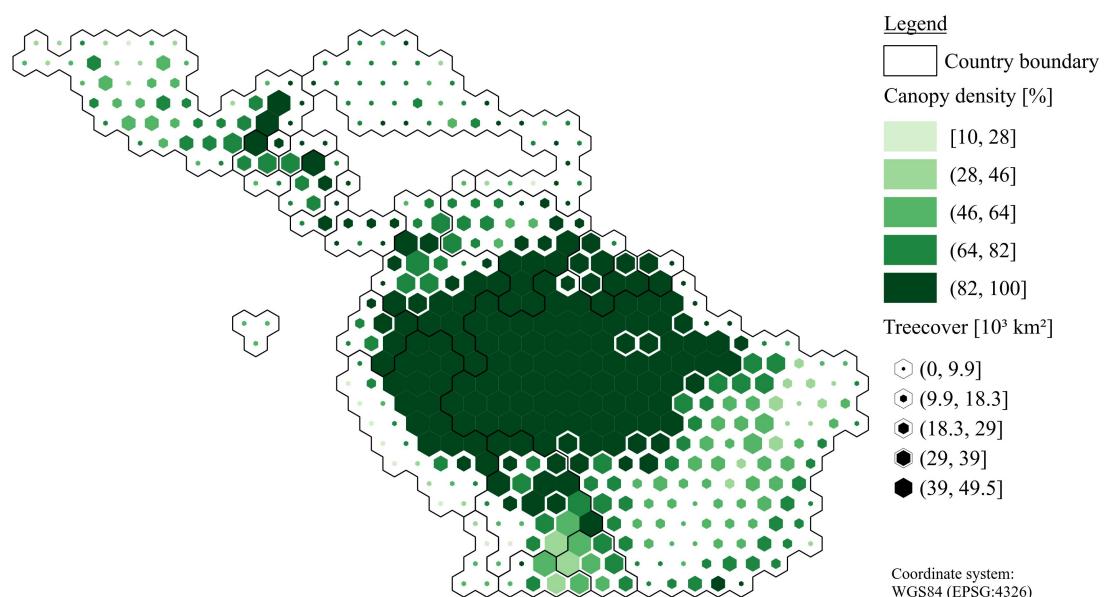


Figure 3.2.

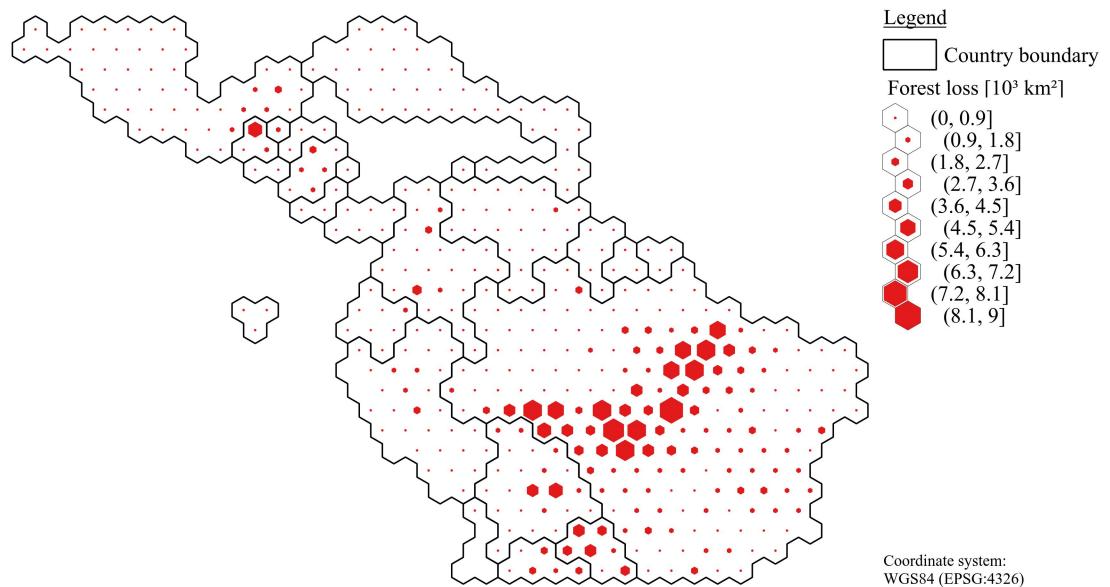


Figure 3.3.

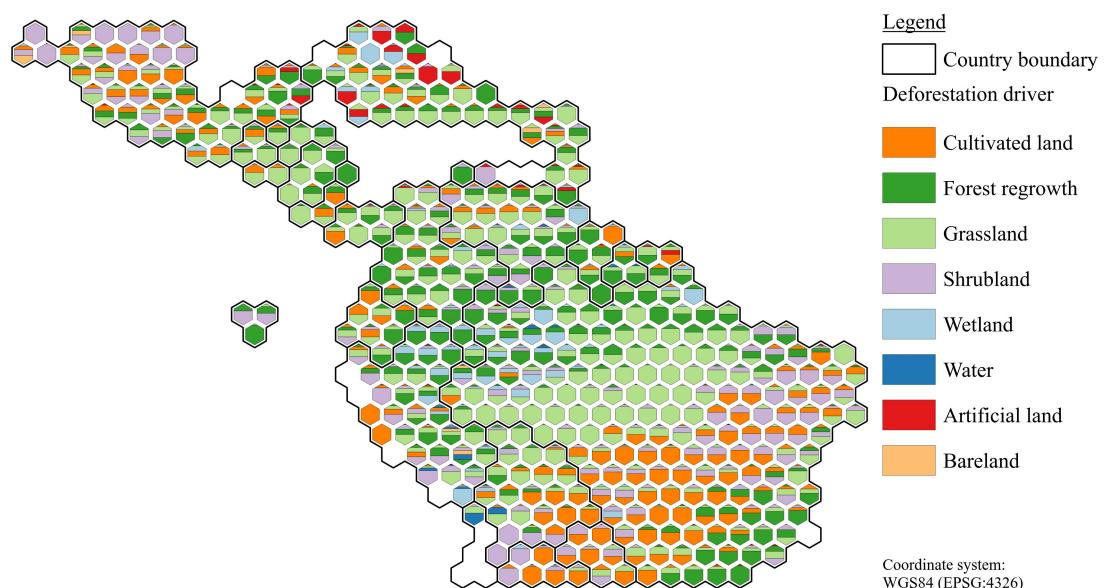


Figure 3.4.

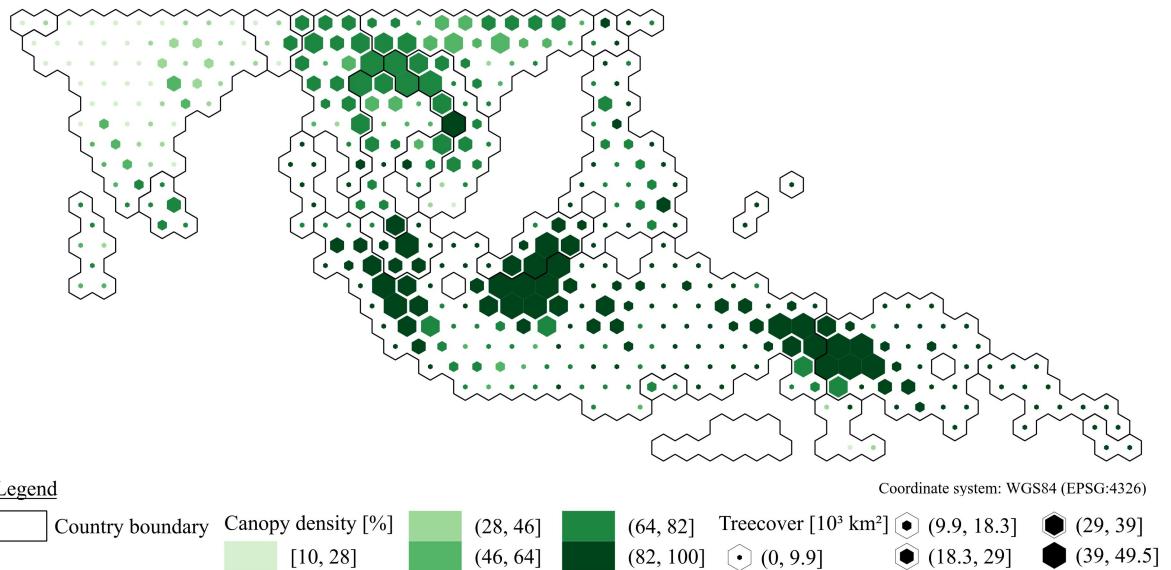


Figure 3.5.

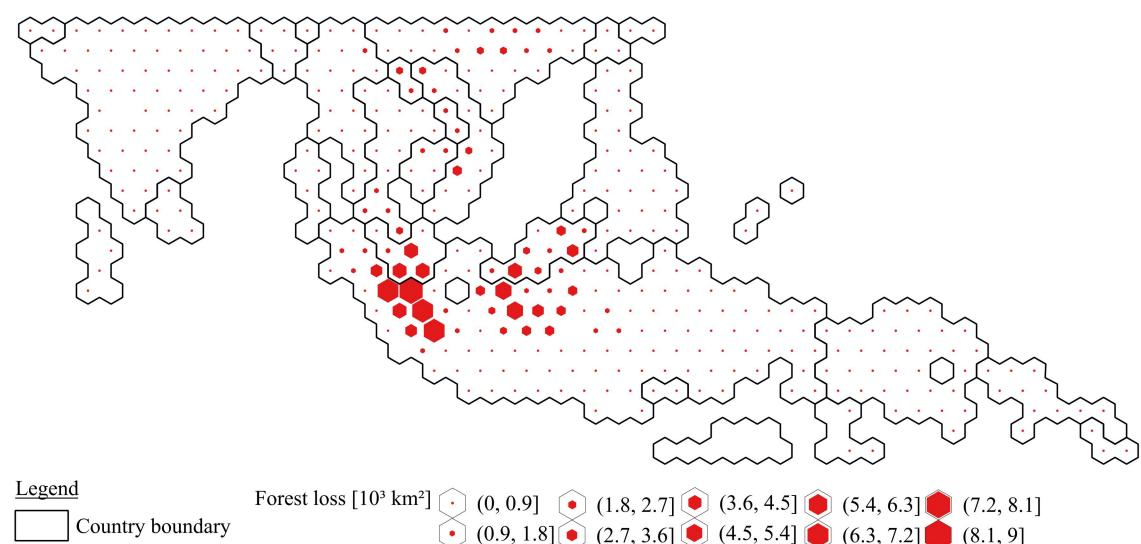


Figure 3.6.

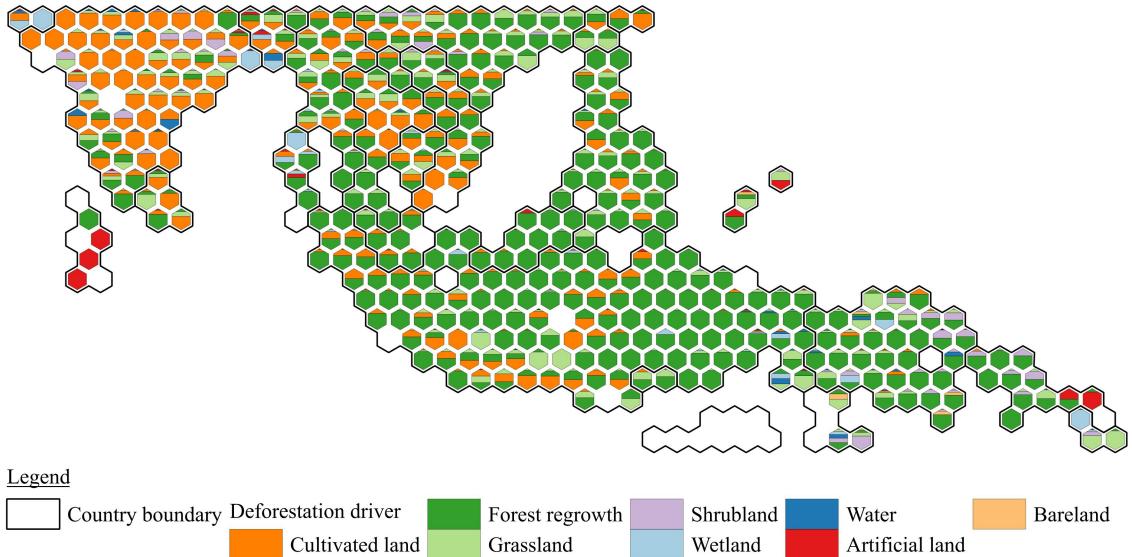


Figure 3.7.

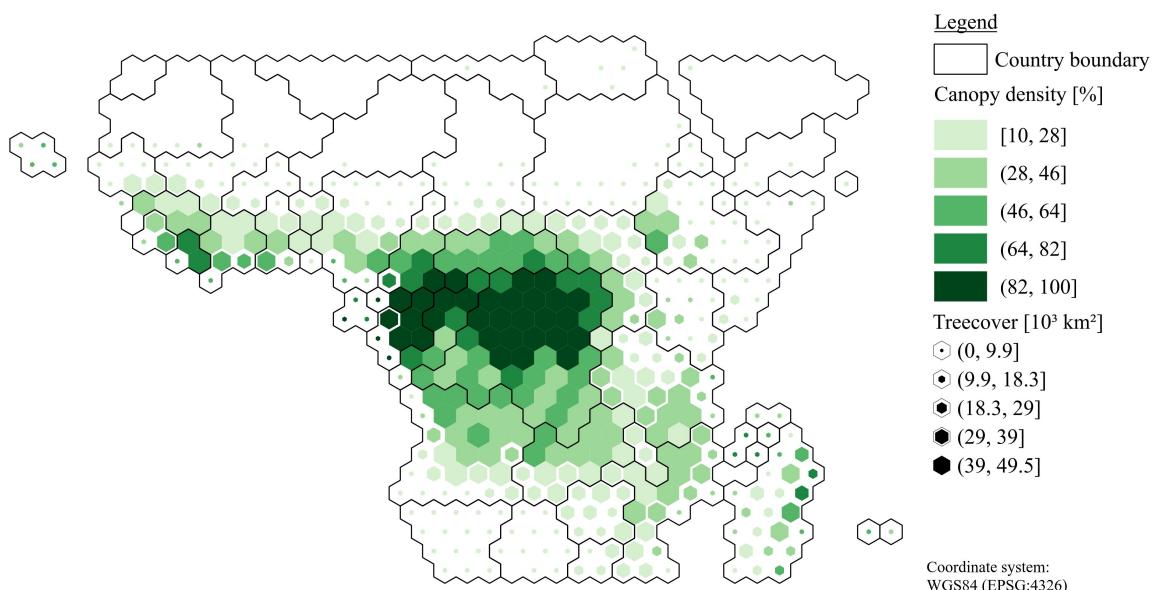


Figure 3.8.

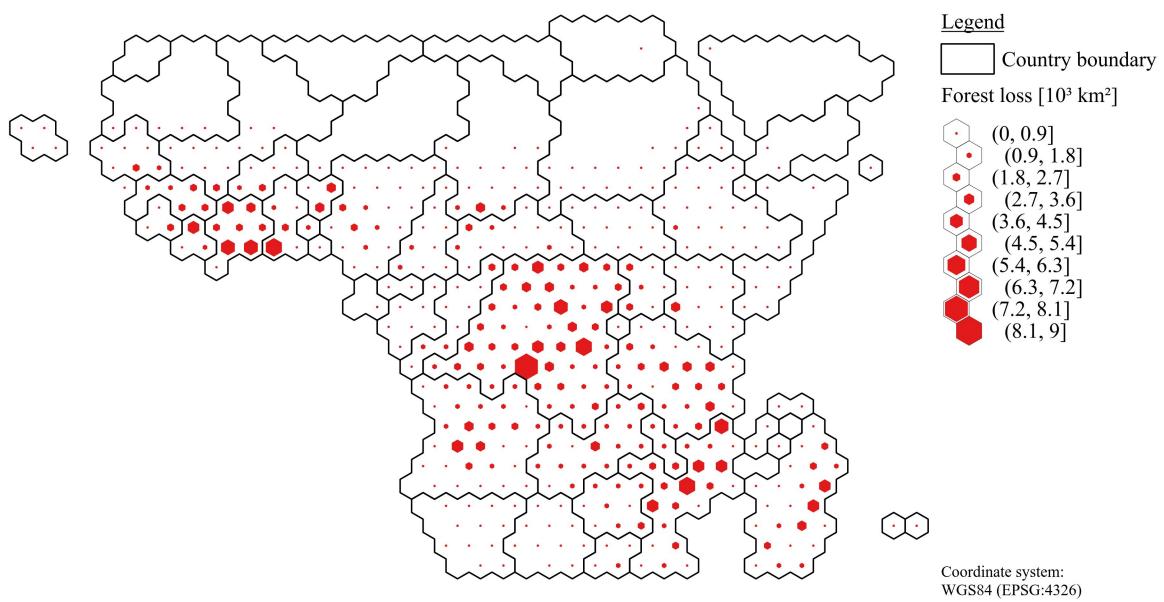


Figure 3.9.

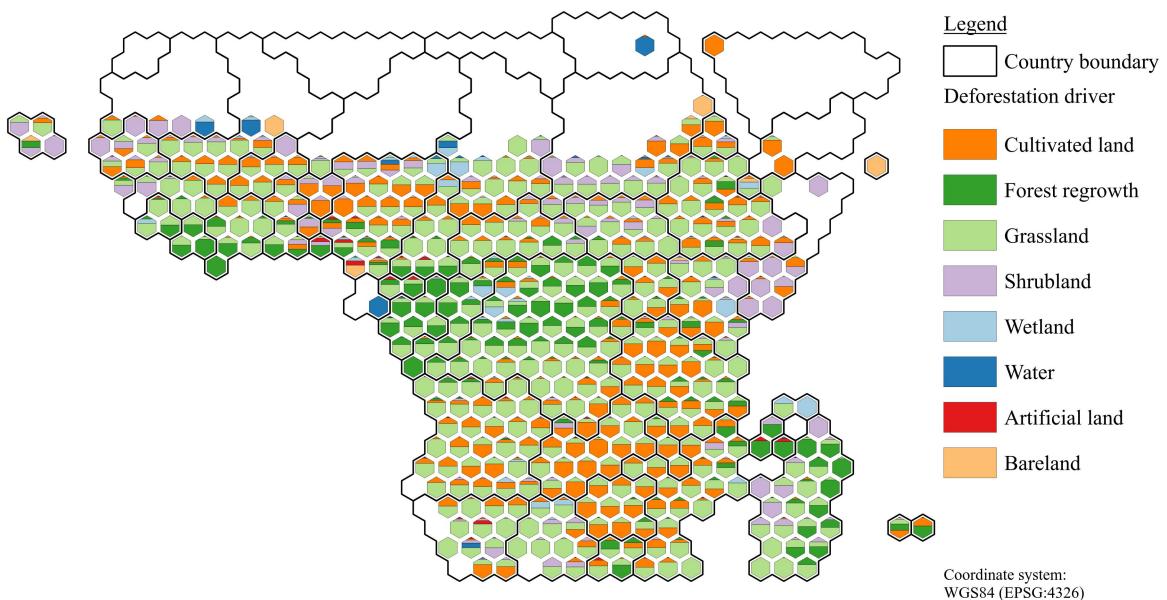


Figure 3.10.

3.3 Deforestation

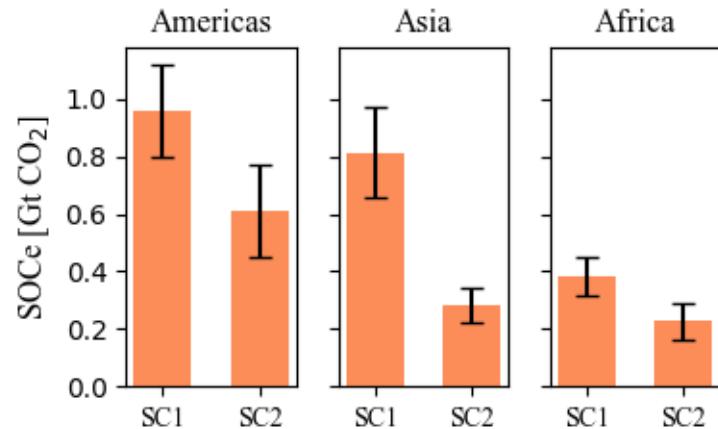


Figure 3.12.

Table 3.5. Soil organic carbon emissions

Region	SC1 [Gt CO ₂]			SC2 [Gt CO ₂]			SC3 [Gt CO ₂]		
	min	mean	max	min	mean	max	min	mean	max
Americas	0.80	0.96	1.12	0.45	0.61	0.77	0.43	0.59	0.76
Asia	0.66	0.81	0.97	0.22	0.28	0.34	0.22	0.28	0.33
Africa	0.32	0.39	0.45	0.17	0.23	0.29	0.16	0.23	0.29

3.3.1 Global

3.3.2 Americas

3.3.3 Asia

3.3.4 Africa

3.4 Ecosystem service value balance

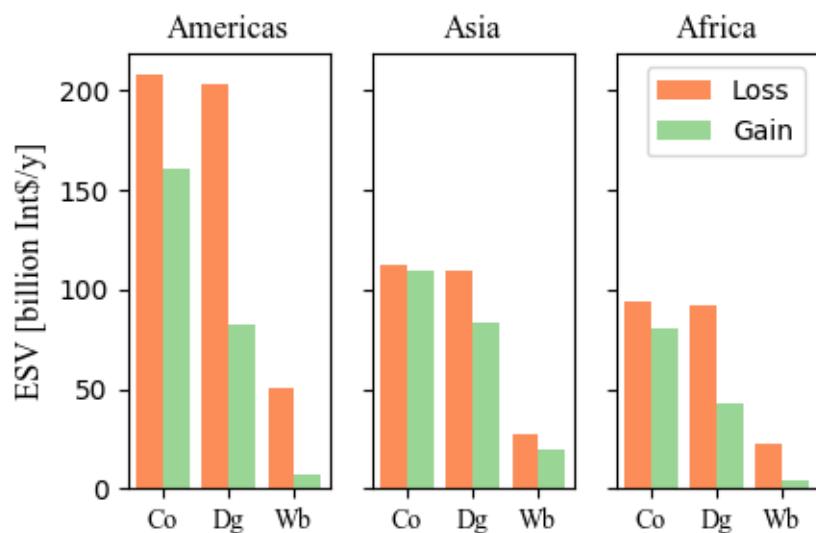


Figure 3.13.

3.4.1 Global

3.4.2 Americas

3.4.3 Asia

3.4.4 Africa

4 Discussion

- Reclassification is not a good idea cuase this approach leads not to consistent results
- Cut polygon by line the scala approach explained with parametric separation function, and bezier

Acknowledgements

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List of Abbreviations

AGB	Aboveground live woddy Biomass density
AISM	Aligned Image Stack Mosaic
API	Application Programming Interface
CRS	Coordinate Reference System
CSV	Comma Separated Values
DEM	Digital Elevation Map
DT	Decision Tree
ESV	Ecosystem Service Values
ETM+	Enhanced Thematic Mapper Plus
FAO	Food and Agriculture Organization of the United Nations
FTP	File Transfer Protocol
GADM	Global Administrative Areas Map
GFC	Global Forest Change
GFW	Global Forest Watch
GIS	Geographic Information System
GL30	GlobeLand30
GLAS	Geoscience Laser Altimeter System
GSOCmap	Global Soil Organic Carbon map
GSP	Global Soil Partnership
GTiff	Geo-Tiff
GeoJSON	Geographic JavaScript Object Notation
IFL	Intact Forest Landscapes
IPCC	Intergovernmental Panel on Climate Change
ISRIC	International Soil Reference and Information Center
ITPS	Intergovernmental Technical Panel on Soils
LC	Land Cover
LCC	Land Cover Change
LIDAR	Light Detection and Ranging
LU	Land Use
LUC	Land-use Change
LULC	Land Use/Land Cover
ME	Mean Error
MLC	Maximum Likelihood Classifier
MLR	Multiple Linear Regression
MODIS	Moderate Resolution Imaging Spectroradiometer
NDVI	Normalized Difference Vegetation Index
POK	Pixel-Object-Knowledge

REGEX	Regular Expression
RF	Random Forest
RMSE	Root Mean Square Error
SD	Standard Deviation
SHP	Shapefile
SOC	Soil Organic Carbon
SOCC	Soil Organic Carbon Content
SVM	Support Vector Machine
UN	United Nations
URL	Uniform Resource Locator
UTM	Universal Transverse Mercator
WGS84	World Geodetic System 1984
stdlib	Standard Library

Appendix

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