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## Article

# Mapping Mangrove Zonation Changes in Senegal with Landsat Imagery Using an OBIA Approach Combined with Linear Spectral Unmixing

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**Abstract:** The mangrove areas in Senegal have fluctuated considerably over the last few decades, and it is therefore important to monitor the evolution of forest cover in order to orient and optimise forestry policies. This study presents a method for mapping plant formations to monitor and study changes in zonation within the mangroves of Senegal. Using Landsat ETM+ and Landsat 8 OLI images merged to a 15-m resolution with a pansharpening method, a processing chain that combines an OBIA approach and linear spectral unmixing was developed to detect changes in mangrove zonation through a diachronic analysis. The accuracy of the discriminations was evaluated with kappa indices, which were 0.8 for the Saloum delta and 0.83 for the Casamance estuary. Over the last 20 years, the mangroves of Senegal have increased in surface area. However, the dynamics of zonation differ between the two main mangrove hydrosystems of Senegal. In Casamance, a colonisation process is underway. In the Saloum, *Rhizophora mangle* is undergoing a process of densification in mangroves and appears to reproduce well in both regions. Furthermore, this study confirms, on a regional scale, observations in the literature noting the lack of *Avicennia germinans* reproduction on a local scale. In the long term, these regeneration gaps may prevent the mangrove from colonising the upper tidal zones in the Saloum. Therefore, it would be appropriate to redirect conservation policies towards reforestation efforts in the Saloum rather than in Casamance and to focus these actions on the perpetuation of *Avicennia germinans* rather than *Rhizophora mangle*, which has no difficulty in reproducing. From this perspective, it is necessary to gain a more in-depth understanding of the specific factors that promote the success of *Avicennia germinans* seeding.

**Keywords:** mangrove; Senegal; Landsat; OBIA; unmixing; forest structure; forest management; mathematical morphology



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## 1. Introduction

Monitoring mangrove forests is essential for their management and conservation [1–3]. In particular, knowing the dynamics of a mangrove allows us to estimate the future status of the many ecosystem services that it provides [4,5]. For example, mangrove forests can contribute to climate change mitigation by sequestering carbon [6,7] and have therefore become a major focus of conservation efforts on a global scale [8]. More locally, mangroves provide resources directly for timber harvesting [9–11] and indirectly [12] through their role in the abundance of fishery resources [13–15] or as habitats for wildlife. However, mangroves experienced a significant global decline between 1980 and 2005 [16]. This trend has fortunately slowed down significantly in recent years [17,18] and is being reversed in some regions, particularly in West Africa [19].

In Senegal, the upper tidal zone is flooded during the high tides and is very often left drained. It is therefore subject to strong evaporation, resulting in salt precipitation in the

soil. During the drought period, the saltier water did not allow mangrove to reproduce and survive in these areas. *Rhizophora mangle* and *Avicennia germinans* located in this interface have suffered the most from this climatic deterioration [20]. These mangroves have increased in area since the mid-1990s [21,22], and the rainfall recovery has led to a decrease in their salinity levels [20,23]. However, the recovery of overall annual rainfall since the mid-1990s is still associated with high interannual variability [24]. Several studies have highlighted the complexity of the response of Senegalese hydrological systems to recent climate fluctuations [23,25]. Even in the rainy season, the Saloum remains an inverse estuary, while Casamance has undergone changes in its maximum salinity due to upstream freshwater inflows [23]. Other studies have highlighted the varying responses of terrestrial vegetation to climate fluctuations in the Sahelo-Sudanian region [26–28]. The interface between the mangrove forest and the “tannes” (local term for salt flats) fluctuates over time in response to variations in the hydrosystem as the climate changes. Fluctuations in the mangrove–salt flats boundary therefore constitute a range of metastability and resilience [29] of the mangrove ecosystem in Senegalese estuaries [20].

This metastability, which is difficult to detect in studies at the local scale only, has probably contributed to the perception that the mangroves are degrading, whereas mapping by remote sensing has revealed an increasing trend in the area [19,21,22,30]. This perception of mangrove degradation persists in the grey literature, the media and, sometimes, the scientific literature and forms the basis of most environmental policies. In response to these reports of degradation, conservation policies in general and reforestation actions, in particular, have been implemented, mainly by NGOs but with some local initiatives. However, these actions have been seriously questioned and are considered a form of land grabbing [31]. In addition to the scientific controversies surrounding the legitimacy of mangrove reforestation in Senegal [32], there are also contradictions between the global climate change mitigation objectives of REDD+ programs and the local needs of populations. In parallel to reforestation, mangrove conservation measures have been reinforced since the 1970s [31]. These are not always accepted, preventing a consensus between government stakeholders and local populations [33], which can jeopardise the support for the sustainable management of this resource.

Moreover, the increasing trend in mangrove areas in Senegal [19,21,22,30] raises the question of the necessity of stronger conservation measures. The need for such actions is particularly uncertain in light of the increasing surface area of mangroves in Senegal due to changes in rainfall and the significant occurrence of spontaneous regeneration, as these findings testify to their resilience [20].

Furthermore, with this increase in surface area, the species that constitute mangrove plant formations have been observed to have contrasting responses [34]. These observations show different species dynamics between the Saloum delta and the Casamance estuary, raising questions about the effectiveness of an identical conservation policy for the two hydrosystems. This is currently the case with reforestation based solely on the genus *Rhizophora* ssp. [32].

Observations made on the local scale [20,34] indicate the lack of *Avicennia germinans* reproduction in the Saloum, which only participates in the dynamics of the area expansion through stump sprouting of altered but still present stumps from the last drought period [20], showing that *Avicennia germinans* can live but cannot reproduce in the current salinity conditions [34]. In contrast, *Rhizophora mangle* colonises mudflats by seeding. In Casamance, the two species have a balanced age pyramid, which suggests that they both contribute to the dynamics of surface expansion [34]. *Rhizophora racemosa* completes the inventory of dominant species in the Senegalese mangrove. Located at the riverine, it is a relatively stable species in the two hydrosystems. Consequently, it is essential to monitor these developments at the fine scale of estuaries. Therefore, this paper hypothesises that these fine-scale processes participate in the spatial organisation of plant formations, which can be mapped on a regional scale. This mapping should make it possible to quantify

the dynamics of transitions between plant formations, which would then corroborate (or not), at the regional scale, the observations made previously at the local scale.

A simple mapping of mangrove surfaces alone is not sufficient to estimate the spatiotemporal dynamics of plant formations in Senegalese mangroves. Plant formations are an estimate of the species composition of the environment. In order to effectively adjust environmental policies, it is important to precisely characterise the dynamics specific to each region to optimise conservation strategies.

Several remote sensing methods have been developed to map mangrove species or plant formations. High-resolution images, such as those from the Landsat satellite, are mainly used to map mangroves alone [20,30,35,36]. Fewer studies have used Sentinel-2 images, although they are effective for vegetation mapping. Very-high-resolution imagery can be used to identify mangrove species [37–39]. However, the cost of very-high-resolution images does not allow for regional-scale mapping, and they remain financially inaccessible in low- and middle-income countries with mangroves.

In a comparison of Landsat OLI-8, Sentinel-2 and Pleiades-1 data, a previous study [37] showed that Landsat images were less effective than Sentinel-2 images for mapping mangrove plant communities, with Pleiades being the most effective. Nevertheless, the authors highlighted the high potential of Sentinel-2 images at a 10-m resolution for mapping mangrove plant communities. The mapping of three mangrove plant formations in Senegal with an OBIA approach using Sentinel-2 images [40] produced interesting results that corroborated the findings of reference [37]. The sizes of the objects that represent vegetation formations were also quantified in this study and were found to be slightly smaller than 60 m, on average. Therefore, this suggests the potential effectiveness of Landsat images merged to a 15-m resolution with an OBIA approach, where the gradient image calculations can be performed with a  $3 \times 3$  pixels window.

Currently, only Landsat images offer a temporal resolution that allows an assessment of mangrove evolution on decadal scales, enabling the study of the resilience of mangroves since the rainfall recovery in Senegal. Furthermore, since the launch of Landsat ETM+ in 1999, the ability to obtain imagery in the panchromatic band has made it possible to achieve a resolution of 15 m using pansharpening methods. In addition, studies have shown that, following certain fusion methods, the NDVI can retain a largely acceptable spatial structure and coherence of spectral values [41], even if the NIR wavelength range does not cover the panchromatic range. NIR or the derived indices often significantly improve mangrove discrimination and are essential for detecting canopy densities [42–44] or different vegetation formations [40]. Therefore, the objective of this study was to assess changes in mangrove zonation from 2000 to 2018 using 15-m Landsat imagery. The results fill a current knowledge gap on this subject and can facilitate policymaking.

Several methods with high-resolution images have been used to map mangrove plant formations or species. OBIA approaches often obtain the best results [45,46]. Pixel-based approaches with Landsat imagery and the use of Tasseled Cap indices were shown to be effective in a diachronic assessment of mangrove zonation dynamics in the Mekong Delta [47]. Linear spectral unmixing was shown to be very useful for mapping the degree of mangrove canopy closure using Sentinel-2 imagery [46] and allowed for a more refined assessment of the degradation and growth dynamics of the surface. In Taureau et al. (2019) [38], Fully Constrained Linear Spectral Unmixing allowed for a standardised comparison between locations and showed a good fit with species zonation patterns in very-high-resolution images. Furthermore, this study reported that the variation in plant proportions between formations dominated by *Sonneratia* spp., *Rhizophora* spp. or *Avicennia* spp. could enable the assessment of the species composition of mangrove formations [38]. Gudex-Cross et al. (2017) [48] were able to improve the discrimination of mixed forest canopy using Landsat images and a combination of linear unmixing and an OBIA approach. In this way, the detection of local differences in vegetation formations, which are sometimes heterogeneous, is envisaged in this work. Therefore, this paper hypothesises that Linear Spectral Unmixing (LSU) calculated from Landsat images merged to a 15-m resolution can

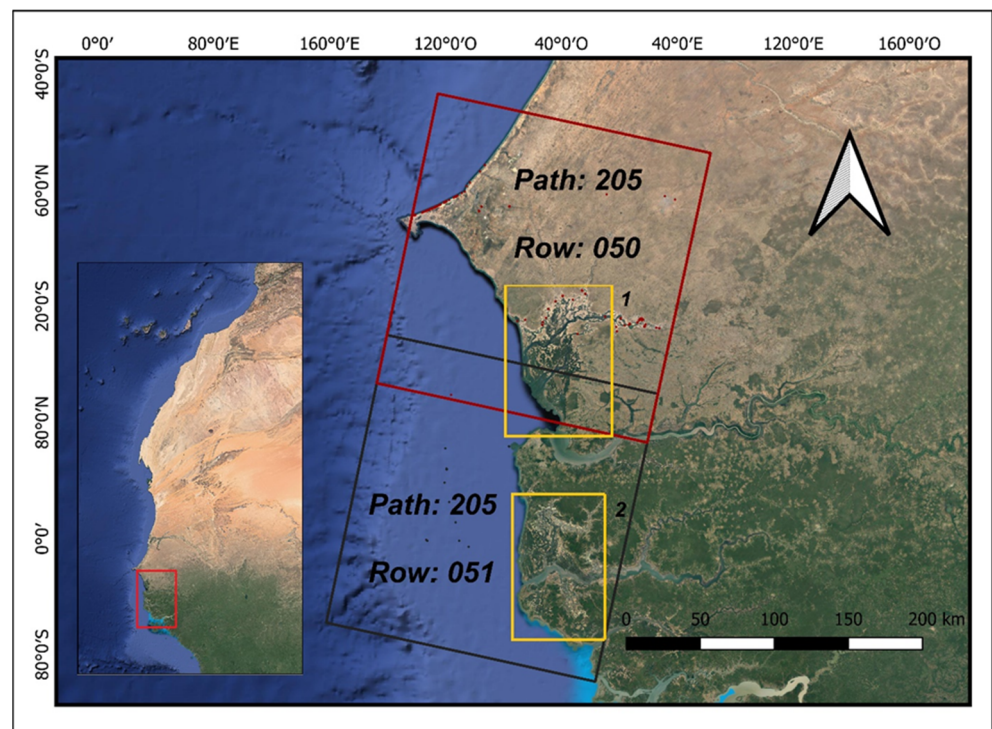


discriminate the degree of openness of the Senegalese mangrove canopy and, thus, enable the mapping of different patches that represent different plant formations using an OBIA approach. Ultimately, the aim of this study was to assess, on a regional scale, the changes in three plant formations resulting from processes on a finer scale, which should make it possible to understand the spatial effects of different biological processes at work for each species. Determining the dynamics could allow us to assign priorities to actions in policies and restoration programs that target particular regions (Saloum or Casamance) and species (*Rhizophora* spp. or *Avicennia germinans*).

## 2. Materials and Methods

### 2.1. Study Sites

The Saloum, located in the north, is a mangrove delta in the Sahelian zone (Figure 1). The rainfall is around 600 mm in the north and 800 mm in the south in the years that have average climate conditions over a century [49]. The river input into the delta is almost nil all year round, giving it an inverse function in which the salinity increases from downstream to upstream, even during the rainy period. Casamance is a mangrove estuary located in the Sudanian zone (Figure 1). The cumulative rainfall can reach 1300 mm in the years that have average climatic conditions over a century. Casamance also is an inverse estuary. However, during the rainy season, the river's contribution can reverse this trend. The highest salinity values move 50–80 km downstream [23].



**Figure 1.** Location of mangroves in the Saloum delta (1) and Casamance estuary (2) and areas of study. The red and black rectangles represent the spatial extent of the Landsat. The yellow rectangles represent the spatial extent of the two mangrove areas.

Three mangrove species are locally dominant in these two hydrosystems: *Rhizophora racemosa* resides at the lowest and most frequently flooded point of the tidal zone, *Avicennia germinans* occupies the middle to upper tidal zone, where the duration of flooding is the shortest, and *Rhizophora mangle* is located at the intermediary between these two ecological conditions. Although they vary in their respective abundance, these three species constitute the mangrove plant formations in Senegal and can be separated into three relatively homogeneous entities. The edges of channels or mudflats that are very frequently submerged are

occupied by a dense, high formation made up solely of *Rhizophora* spp. and dominated by *Rhizophora racemosa*. Then, a low and dense mangrove is dominated by *Rhizophora mangle*. Finally, in the rarely flooded tidal zone, an open formation with a lower plant fraction contains a mixture of *Rhizophora mangle* and *Avicennia germinans*. A gradient of spectral values is regularly observed locally, which, depending on the indices, seems to reflect this zonation to some extent.

## 2.2. Satellite Data Acquisition and Preprocessing

The study area is covered by two Landsat image tiles obtained through the US Geological Survey (USGS) with referencing paths of 205 for both regions and rows of 50 and 51 for Saloum and Casamance, respectively. The surface reflectance images are from the USGS Collection 2 Level 2 and were produced using the LaSRC and LEDAPS algorithms for the Landsat 8 OLI and Landsat 7 ETM+ data, respectively (Table 1). Both image sets were corrected for atmospheric effects. Panchromatic images of the area were acquired on the same dates and were taken from Collection 2 Level 1, because they were not included in Collection 2 Level 2. Two Landsat 8 OLI images were acquired on 18 December 2018 for the end date of the diachronic study, and two Landsat ETM+ images were acquired on 8 December 2000 for the initial date. The month of December was chosen, because it allowed us to work with images in the absence of cloud cover and in temporal proximity to the rainy season (from June to September). There is a gradient in the reflectance of the NIR (Near infrared: 0.775–0.90  $\mu\text{m}$ ) from east to west, which increases as the dry season progresses and provides a finer discrimination of mangrove vegetation formations. The images were mosaicked and then cut to obtain the two distinct and complete mangrove estuaries.

**Table 1.** Summary of the sensors and products used in the study and their corresponding characteristics.

Dates of Acquisition by Path/Row					
205/50 and 205/51	Capture	Collection and Level of Product	Spatial Resolution	Bands	Spectral Range( $\mu\text{m}$ )
08/12/2000	ETM+	C2/L2	30 m	1	0.45–0.515
		C2/L2	30 m	2	0.525–0.605
		C2/L2	30 m	3	0.63–0.69
		C2/L2	30 m	4	0.775–0.90
		C2/L2	30 m	5	1.55–1.75
		C2/L2	30 m	7	2.08–2.35
		C2/L1	15 m	8	0.52–0.9
		18/12/2018	OLI	C2/L2	30 m
C2/L2	30 m			3	0.525–0.600
C2/L2	30 m			4	0.630–0.680
C2/L2	30 m			5	0.845–0.885
C2/L2	30 m			6	1.560–1.660
C2/L2	30 m			7	2.100–2.300
C2/L1	15 m			8	0.500–0.680

## 2.3. Pansharpening

The mangrove vegetation formations are patches, each with its own morphological characteristics. The high and dense mangrove, often located on the edge of the channel, is a long and narrow strip of vegetation. It rarely exceeds several dozens of metres in depth on the foreshore. From this point of view, the Landsat data at a 30-m resolution are too coarse to discriminate the patches of high and dense mangrove on the edge of the channel with an OBIA approach. However, an OBIA approach using Sentinel-2 images at 10 m has shown some robustness [40]. A compromise was therefore found by using pansharpening fusion method to obtain images at a 15-m resolution.

The IHS (Intensity–Hue–Saturation) method was chosen for its speed of implementation and the good results that it generates in subsequent calculations of spectral indices,

such as the NDVI [41]. In addition, it was important to use a method that incorporates spectral and spatial information by adding rather than multiplying for the calculation of normalised spectral indices [41]. Although the panchromatic band does not cover such a large spectral range, pansharpening was carried out over all spectral bands of the Landsat 7 ETM+ and Landsat 8 OLI.

#### 2.4. Isolating the Mangrove

An OBIA approach solely on mangrove pixels was used in this study in order to maximise the spectral contrast between the mangrove formations. To this end, unsupervised classification was used to group the images into five classes: water, mangrove, mudflats, unwooded dryland and wooded dryland. Water and mudflats were designated as separate classes, because they conditioned the sampling of the segmentation regions and were used for spectral demixing. The classification of wooded and non-wooded land within the mainland made it possible to separate certain pixels that did not present a strong contrast between the mudflats and dense mangrove, respectively. Four spectral indices were created, which accelerated the classification process of the different masks due to their contrasting spectral responses to the different discriminated surface states (Figure 2). The NDVI (Normalized Difference Vegetation Index), NDWI (Normalized Difference Water Index), MNDWI (Modified Normalized Difference Water Index) and NDBI (Normalized Difference Built Index) were calculated with the following equations:

$$NDVI = \frac{(NIR - R)}{(NIR + R)}$$

$$NDWI = \frac{(NIR - SWIR)}{(NIR + SWIR)}$$

$$MNDWI = \frac{(Green - SWIR)}{(Green + SWIR)}$$

$$NDBI = \frac{(SWIR - NIR)}{(SWIR + NIR)}$$

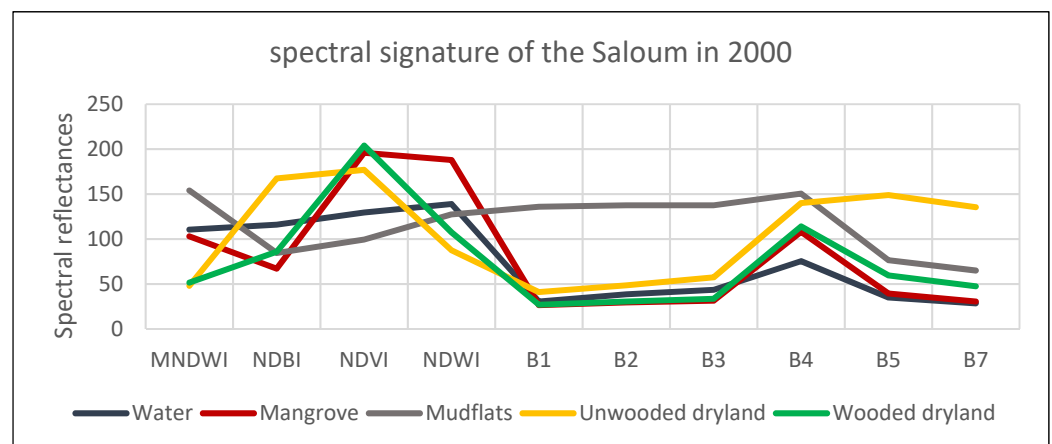
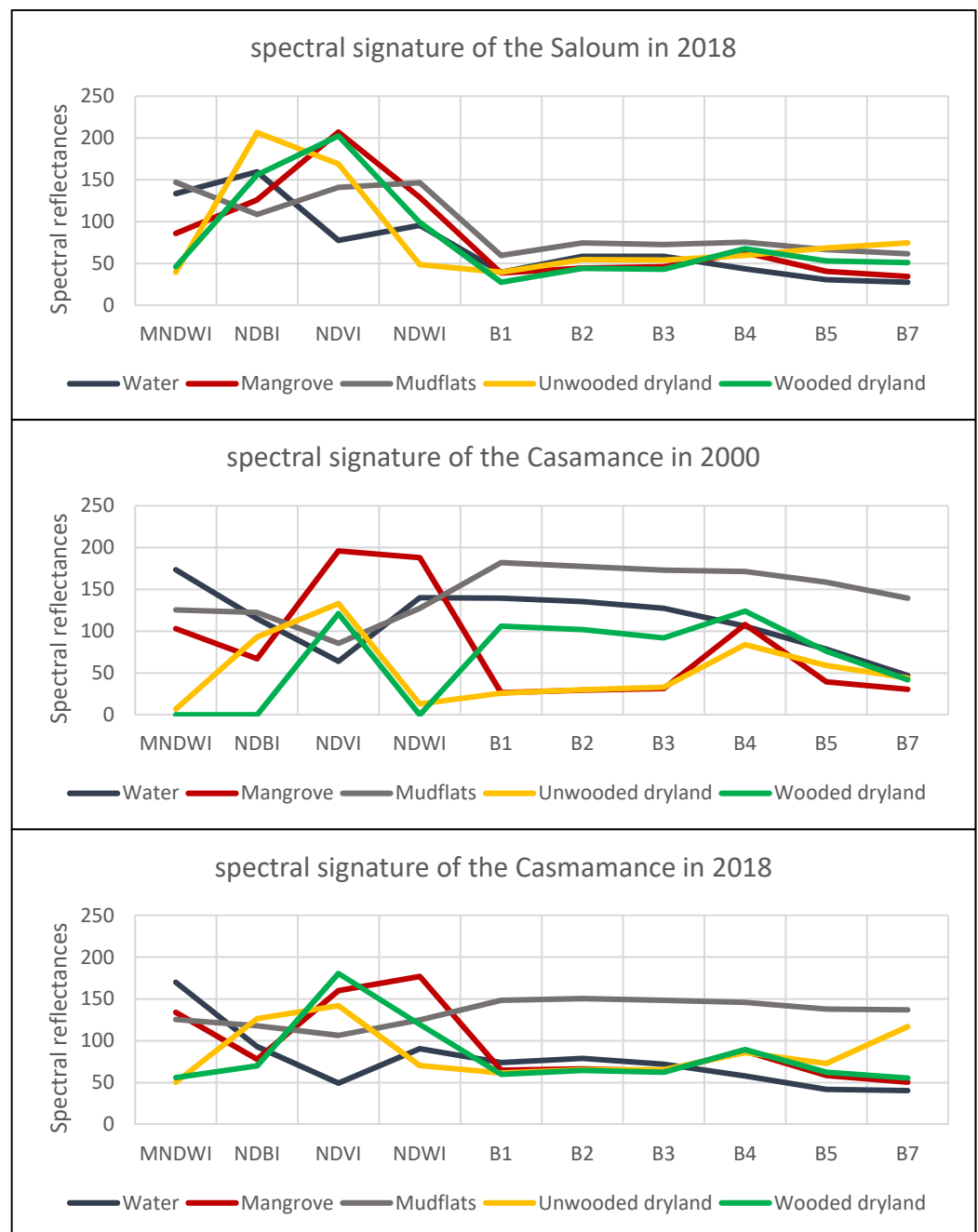


Figure 2. Cont.



**Figure 2.** Spectral signature of the five land cover classes. The spectral signatures were calculated on each Landsat scene, on both dates and on each area. The values on the y-axis represent the average reflectance for each class.

### 2.5. Stacked Unsupervised Classification

The mangroves in the Saloum and Casamance were mapped using unsupervised classification with the ISODATA algorithm. The discrimination of mangrove surfaces was optimised using stacked classifications [50]. The spectral index allowed a relatively precise initial classification and the extraction of the initial masks. The masks were used as the basis of principal component analysis, allowing for increased contrasts and discrimination by a second ISODATA classification of the components. These “subclasses” were also analysed with the spectral curve to confirm that they were classified into the correct land use type, and if necessary, they were reallocated in a reclassification process. The results of the classifications were used to extract a binary image (mask) of the mangrove surfaces,

which was used to calculate the vegetation fraction and to implement the OBIA approach using these pixels alone.

### 2.6. Linear Spectral Unmixing and Plant Fraction

The spectral signatures and indices calculated for the mangrove mask highlight local differences that reflect the zonation of mangrove plant formations. The NDVI proved to be an efficient way to represent this distribution. Beyond its local structure, a longitudinal regional structure was observed. Locally, for each plant formation, the NDVI decreased along the tidal zones. The physiognomy of the vegetation changed from east to west and influenced the NDVI values. Therefore, two different vegetation formations can have the same NDVI value depending on their location within the estuary. Furthermore, as mentioned by references [38,46], the NDVI is strongly dependent on atmospheric and weather conditions prior to imaging, which can bias the discrimination of plant formations. These two studies obtained the best correlation of the vegetation fraction by performing linear spectral unmixing (LSU) and canopy closure calculation. The vegetation fraction is derived from the decomposition of the pixels and their fractional cover of each extremity member. The typology of the mangrove plant formations chosen here has a component that depends on the degree of closure and justifies the choice for LSU. Furthermore, the values obtained for the vegetation fraction are less dependent on the environmental conditions and allow for greater reproducibility over time.

To obtain the plant fraction and quantify the degree of mangrove canopy closure, two steps were necessary. First, representative pure spectra were selected as endmembers e.g., non-turbid dark water, salt flats with maximum moisture characteristics and mangrove areas of maximum photosynthetic activity. Then, the reflectance of the pixels was decomposed according to these spectra to quantify the proportion of endmembers in each pixel.

### 2.7. Endmember Selection

Through endmember selection, the LSU assumes that the values of pixels can be modelled by the linear combination of different land cover spectra. Here, it is assumed that a vegetation formation can be discriminated by its vegetation fraction, which, in mangroves, is related to its water and mud fractions. Three endmembers were therefore retained: water, mangrove and mudflats. Two difficulties had to be considered in their selection. First, the endmembers must have radiometric characteristics at the extremes of the pure spectra, and second, the indicators used to select them must be simple in order to ensure their reproducibility over time. From this point of view, the pure spectra of vegetation were identified according to the maximum NDVI value of the mangrove pixels. The maximum NDWI value, calculated for the mask, was obtained for the pure mudflat spectra. The gradient of NDWI values on mudflats is closely related to their submersion duration [51]. Finally, a six-class classification of the water mask was performed, and pixels with spectral signatures with the lowest mean reflectance were extracted for the pure water spectrum. These pixels correspond to very deep water with no signs of turbidity.

### 2.8. OBIA

The use of an OBIA approach is therefore necessary to discriminate the three mangrove plant formations: zonation of the three formations is present in the east and west of the Saloum and Casamance, whereas the local gradients are not always as pronounced along the longitudinal axis. These slight local variations cannot be discriminated by pixel approaches, whereas OBIA approaches using image segmentation methods can distinguish between these different objects.

The Watershed algorithm of Terrset 18.21 software was chosen to perform segmentation for two reasons. First, mangrove vegetation varies along the tidal zone and environmental gradients. Therefore, a contour-based approach can be used to quantify these transitions via a gradient image and to thereby obtain data that inform the choice of region



merging thresholds. Second, region-based approaches are very effective for highly textured, high-resolution images in which transitions in luminance values are very progressive, which is not the case with images at a resolution of 15 m.

The images from the LSU vegetation fraction calculation show much more pronounced local contrasts than the NDVI. This leads to very little change in the number of segments generated at different scale parameters. The Watershed algorithm first calculates the gradient image (the variance image in the case of IDRISI Terrset) and then delineates the regions from the local minima, where the dividing line is formed by high variance values. The algorithm makes it possible to merge the regions by applying a tolerance criterion. Here, no merging was carried out because the contrasts made it possible to obtain objects that were sufficiently dissimilar to reveal distinct vegetation formations, thus limiting the constraints imposed by a resolution of 15 m.

The regions were then selected to train a supervised classification algorithm. The KNN (k-nearest neighbour algorithm) was used in this work. KNN obtained better kappa than Random Forest or Maximum Likelihood classifiers. It was therefore chosen for the classification.

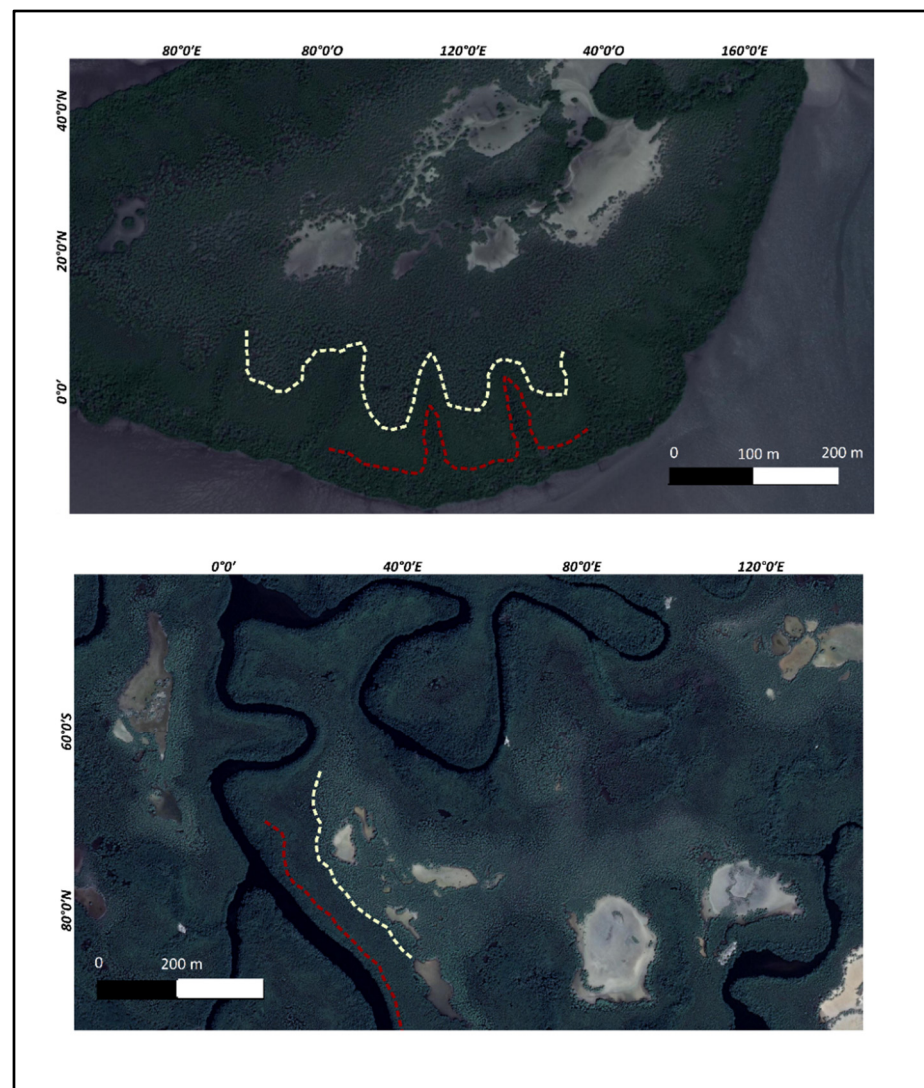
The classification of plant formations was evaluated by selecting regions and comparing them with very-high-resolution Google Earth (GE) image.

In this work, we identified mangrove plant formations as follows:

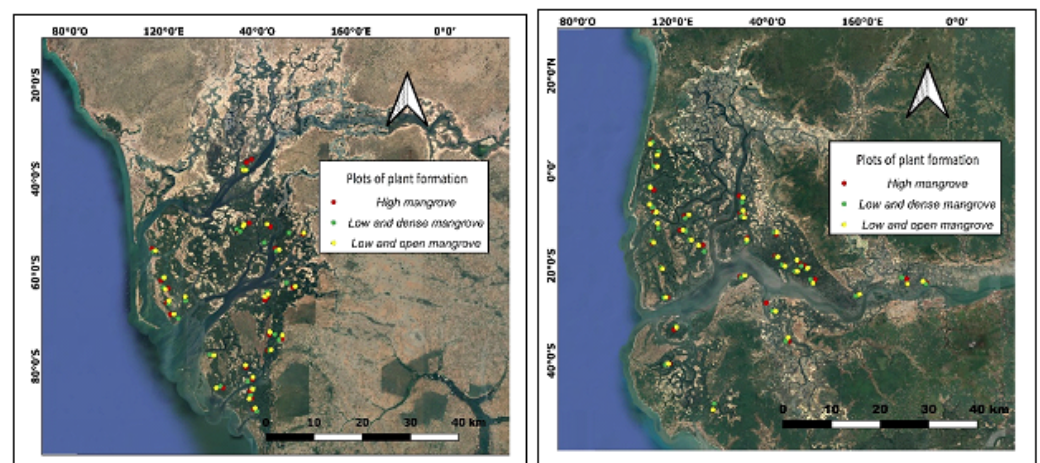
- High riverine mangrove dominated by *Rhizophora racemose*: High mangrove.
- Low and dense mangrove dominated by *Rhizophora mangle*: Low and dense mangrove.
- Low and open mangrove with mixed *Rhizophora mangle* and *Avicennia germinans*: Low and open mangrove.

Mangrove plant formations have different characteristics, mainly related to the height, degree of canopy closure and heterogeneity of plant cover and colour. These attributes are identifiable in a very-high-resolution image and allow for photointerpretation (Figure 3). However, the lack of very-high-resolution images for the year 2000 prevented the comparison and validation of the 2000 maps. The kappa index could only be calculated for images of 2018 for comparison with Google Earth (GE) image.

To overcome this obstacle and select training plots for the supervised image classification for the year 2000, stable regions between 2000 and 2018 were selected. The advantage of the LSU is that it is less sensitive to phenology than the NDVI [52]. Although few studies have tested the seasonal effect of the vegetation fraction index at a single site, we assumed that if a region has not experienced any change in its vegetation cover density, then it is highly unlikely that the vegetation formation had changed. First, the segments with same spatial shape and extension on both scenes need to be found. Second, these areas should not have undergone significant changes in their vegetation cover. Therefore, the difference in the vegetation fraction was calculated to highlight areas of stability that have neither opened nor closed canopies. Segments with a difference in vegetation fraction between  $-0.10$  (negligible decrease) and  $0.10$  (negligible increase) were retained. They can therefore be considered the stable vegetation formations between 2000 and 2018. Finally, within such segments, a random selection was led, but only within those whose vegetation remained identifiable in the 2018 very-high-resolution imagery (Figure 4). We performed a random partition for the training and validation segments of the classification (Table 2).



**Figure 3.** Very-high-resolution Google Earth (GE) image for the photointerpretation of the three vegetation formations. The **top** is a zonation in the Saloum, and the **bottom** is in Casamance. The red line symbolizes the interface between dense high mangrove and dense low mangrove. The white line symbolizes the interface between the low and dense mangrove and the low and open mangrove.

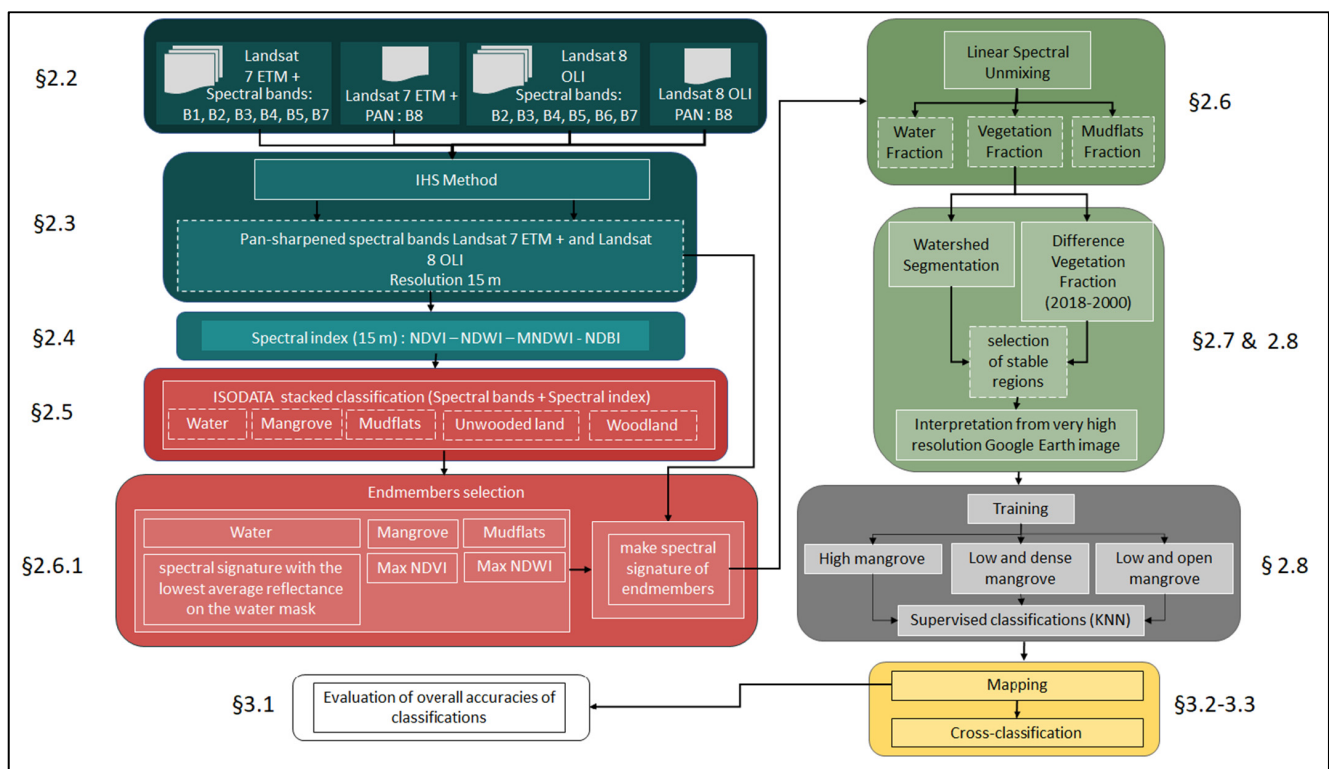


**Figure 4.** Distribution of the validation samples for the mangrove zonation classification (**left**: Saloum; **right**: Casamance).

**Table 2.** Summary of the partition for the training and validation data of the classification (in pixels).

	Saloum		Casamance	
	Training	Control	Training	Control
high mangrove	1109	1055	1812	1335
low and dense mangrove	1827	891	2213	2344
Low and open mangrove	1382	1002	2620	1863
Total (pixel)	7266		12,187	

The flowchart in Figure 5 compiles the different steps of the methodological chain.

**Figure 5.** Methodological flowchart for mapping the mangrove zonation changes.

### 2.9. Mapping of Spatiotemporal Trajectories

The different changes mapped remained difficult to read on a regional-scale map and did not reveal the dynamic processes at work in the mangrove. Consequently, a mapping based on the local frequency and the relative share of each important change in local dynamics was undertaken. Thus, within a 5-km resolution grid, the pixel density of each transition class, as well as its percentage contribution, was calculated.

## 3. Results

### 3.1. Accuracy Assessment

The accuracies for the Saloum (Table 3) and Casamance (Table 4) mangroves are 80% and 83%, respectively (Table 5). According to reference [53], these values indicate good discrimination. The classifications are therefore acceptable in the Saloum and Casamance. These Kappa indices were calculated only on the 2018 validation plots identified on Google Earth (GE) as very-high-resolution images. Nevertheless, the contributions of the vegetation formations to the overall accuracy are different between these two areas. In both cases, the low and dense mangrove accounts for most of the errors. There are 27% errors of

omission in Saloum and 13% in Casamance. The errors of commission are 13% for Saloum and 9% for Casamance. Its interface, in the mid tidal zone, is probably the cause. In Saloum, the major confusion is at its interface with the high and dense mangrove of the shore. However, the low and dense mangrove is mainly confused not with the high mangrove of the channel edge in the Saloum but with the low open and mixed mangrove in Casamance. In Casamance, the errors are equally divided between the two plant formations at the interface. In addition, the open and mixed lowland mangrove obtains almost equivalent discrimination scores on the two maps.

**Table 3.** Confusion matrix for the mangrove zonation classification in Saloum (ground truth in the rows, and categorical images in the columns).

Saloum	High Mangrove	Low and Dense Mangrove	Low and Open Mangrove	Total	Commission
High mangrove	983	236	2	1221	0.19
Low and dense mangrove	58	646	47	751	0.13
Low and open mangrove	14	8	924	946	0.02
Total	1055	891	1002	2948	
Omission	0.06	0.27	0.07		0.13

**Table 4.** Confusion matrix for the mangrove zonation classification in Casamance (ground truth in the rows, and categorical images in the columns).

Casamance	High Mangrove	Low and Dense Mangrove	Low and Open Mangrove	Total	Commission
High mangrove	1202	128	2	1332	0.09
Low and dense mangrove	90	2033	133	2256	0.09
Low and open mangrove	43	177	1696	946	0.11
Total	1335	2344	1863	5542	
Omission	0.09	0.13	0.08		0.11

**Table 5.** Kappa statistics of the mangrove zonation classification.

Overall Accuracy	Kappa
Saloum 2018	0.80
Casamance 2018	0.83

The plots of the ground truths of the changes could hardly be inventoried. In this work, the ground truth plots were collected only in 2018 by the photo interpretation of Google Earth (GE) images. Ground truth plots are derived from the objects detected with the OBIA approach. These are of variable sizes, making it difficult to partition them by proportions. We have chosen to randomly select the objects used for training and validation. The partitioning of the dataset was also done randomly. The purpose of Table 6 is to assess the uncertainty of the results due to the sampling. We followed the recommendations of Olofon et al. (2014) [54] to evaluate the accuracy and confidence intervals of the map classes.

The margins of uncertainty are large for the high and dense mangrove class but remain smaller for the low and dense mangrove class and the low and open mangrove class. The smallest margins of uncertainty are for the classes that are most involved in the change maps developed in this study.



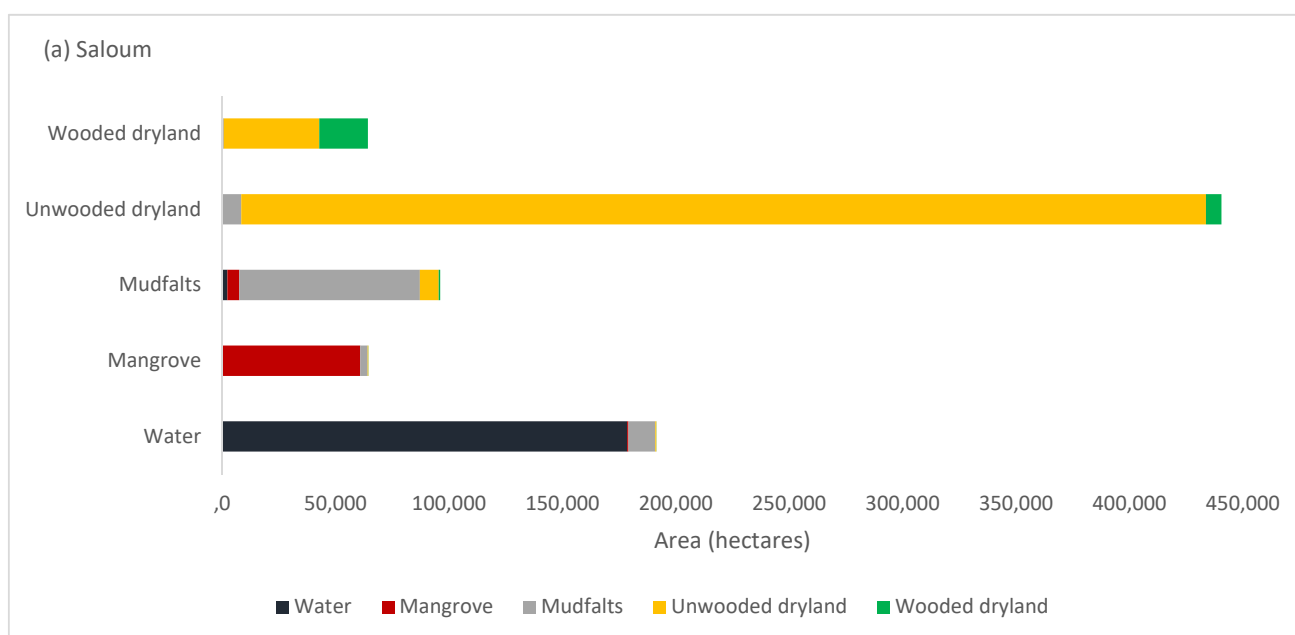
**Table 6.** Differences between the mapped areas and area estimates of the plant formation and relationship to the relative margin of error.

Saloum 2018	Mapped (ha)	Estimated (ha)	Relative Difference	MoE at 95% Confidence
High and dense mangrove	13,549	7603	78%	91%
Low and dense mangrove	30,962	16,829	84%	41%
Low and open mangrove	18,770	21,719	−14%	32%
Casamance 2018	Mapped (ha)	Estimated (ha)	Relative Difference	MoE at 95% Confidence
High and dense mangrove	10,943	10,941	0%	61%
Low and dense mangrove	33,689	24,450	38%	27%
Low and open mangrove	36,221	23,575	54%	28%

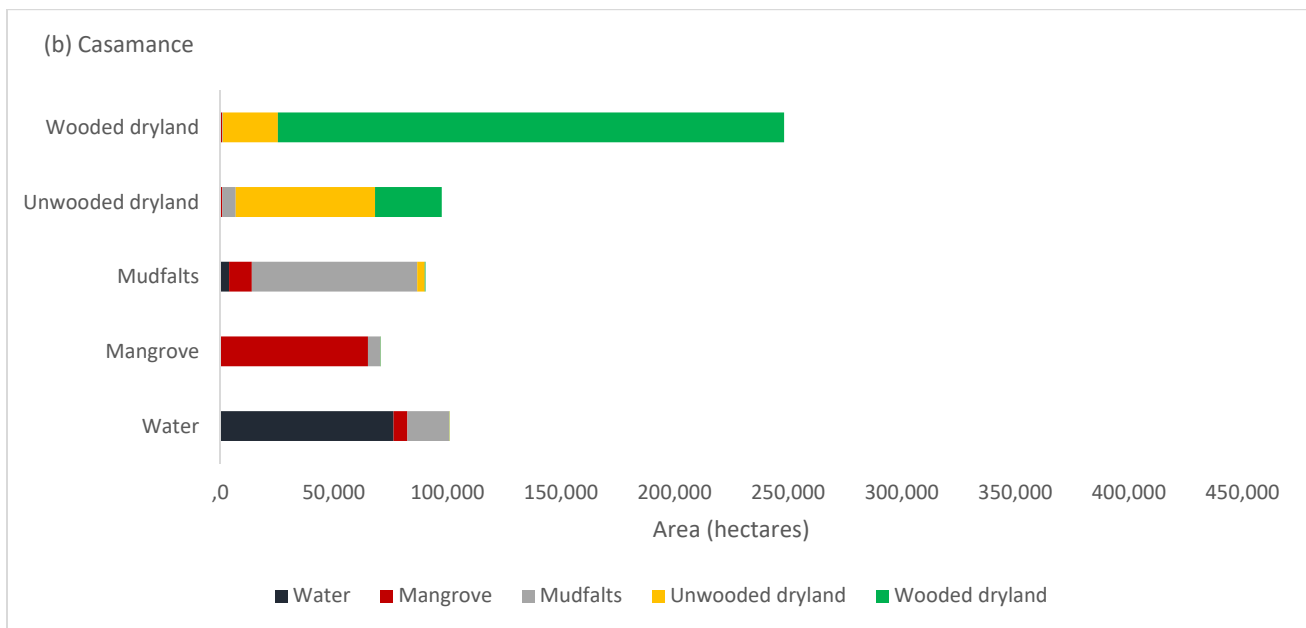
### 3.2. Contribution of Each Formation to Overall Change

#### 3.2.1. Reorganisation of Surfaces

Since the 2000s, mangrove areas have increased in the two main hydrosystems of Senegal. There has been an increase of 5330 hectares of mangrove area in the Saloum delta and 11,614 hectares in the Casamance estuary. These increases are the result of the difference between surface area gains and losses. The Saloum mangrove has experienced a loss of 2996 hectares of its surface area. In Casamance, the loss amounts to 5118 hectares. The rates of change in mangrove surface area between 2000 and 2018 reached 8.4% for the Saloum and 17.25% for Casamance. Of these changes, the colonisation of mudflats between the two dates represented 87% of the increase in surface area in Saloum (Figure 6), while the closure of bolongs, new sedimentations and flooded mudflats contributed to 10% of the new mangrove gains. In Casamance, the proportion of spatial expansion on stable mudflats was calculated to be 56% and 34% for the closure of bolongs (saltwater channel, characteristic of the mangrove areas of Senegal or Gambia) or the colonisation of new sedimentations and flooded mudflats (Figure 6).

**Figure 6.** Cont.

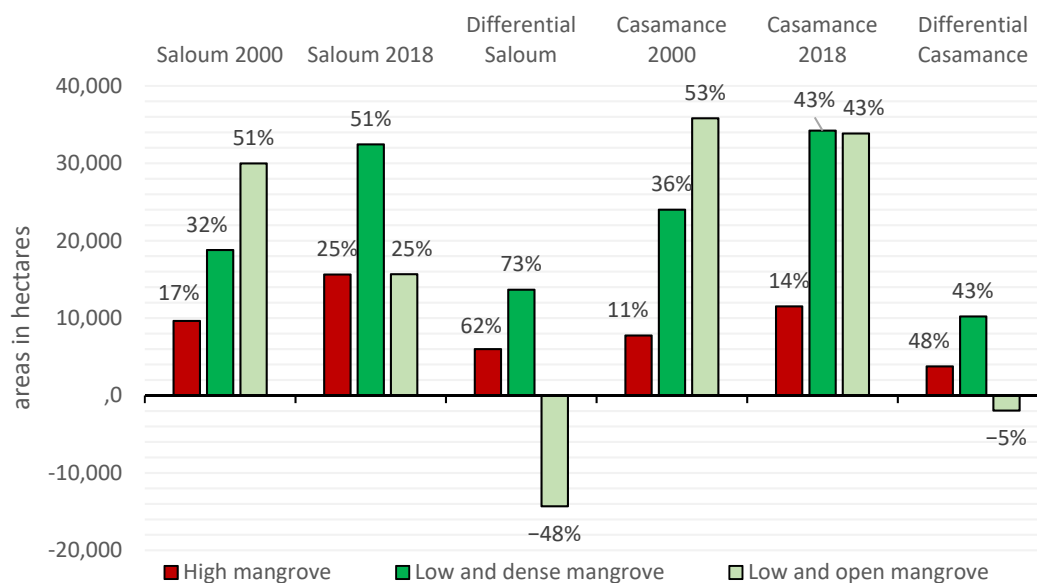




**Figure 6.** Land cover change and contribution of classes to the area change between 2000 and 2018 in Saloum (a) and Casamance (b).

The Saloum mangrove is therefore experiencing a significant increasing trend in mudflats at the interface with the salt flats. This is also the case in Casamance but to a lesser extent. In fact, the colonisation on mudflats located at the lower tidal zone represents a significant proportion of the increase.

Moreover, within the mangroves of Senegal, the modalities of the increases differ. Indeed, the mangrove plant formations do not contribute to the same degree to the changes observed in Casamance and the Saloum (Figure 7). The low and dense mangrove is undergoing major gain in the Saloum. It occupied one-third of the mangrove area in 2000 and now occupies half of it. Conversely, low and open mangrove was the main plant formation in 2000 (51%) and is currently 25% of the mangrove area in the Saloum. In Casamance, the low and dense mangrove has undergone a moderate increase, from 36% to 40% of the mangrove surface.



**Figure 7.** Changes in the surface areas of mangrove formations between 2000 and 2018 in Saloum and Casamance.

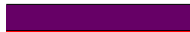
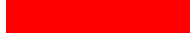







### 3.2.2. Conversion of Surfaces

The matrix of changes provides a more detailed understanding of this spatial reorganisation. The results indicate that the densification of low and open mangrove patches is one of the main factors (Tables 7 and 8) that explain the overall dynamics of change in the regeneration process in both regions. However, this densification is more prevalent in the Saloum, accounting for more than 20% of the observed changes, whereas it amounts to 15% in Casamance. However, the latter also shows much greater gains in low and open mangrove (13.5%) compared to Saloum (6.5%). The opening of the canopy remains low and relatively similar between the mangrove plant formations of the two regions. The opening of the low and dense mangrove is more important in Casamance. A final difference between the two regions involves the stable surface areas of the low and open mangrove. In the mapping of the changes, they are around 15% in Saloum compared to a little more than 20% in Casamance. It is thus necessary to understand the extent to which these stable nuclei have contributed to the dynamics of the spatial expansion of the mangroves in the hydrosystems.

**Table 7.** Transition matrix for the mangrove zonation class in Saloum.

Saloum	No Mangrove	High Mangrove	Low and Dense Mangrove	Low and Open Mangrove	Total
No mangrove		323 ha	426 ha	2248 ha	2996 ha
High mangrove	1410 ha	6246 ha	3925 ha	4050 ha	15,631 ha
Low and dense mangrove	2534 ha	3000 ha	13,148 ha	13,764 ha	32,446 ha
Low and open mangrove	4382 ha	73 ha	1287 ha	9923 ha	15,665 ha
Total	8326 ha	9641 ha	18,786 ha	29,986 ha	

Legend	
	Decrease spatial extension involving high mangrove dominated by <i>Rhizophora racemosa</i>
	Decrease spatial extension involving low and dense mangrove dominated by <i>Rhizophora mangle</i>
	Decrease spatial extension involving low, open and mixed mangrove dominated by <i>Rhizophora mangle</i> et <i>Avicennia germinans</i>
	Opening in vegetation cover
	Increase in spatial extension involving high mangrove dominated by <i>Rhizophora racemosa</i>
	Increase spatial extension involving low and dense mangrove dominated by <i>Rhizophora mangle</i>
	Increase spatial extension involving low, open and mixed mangrove dominated by <i>Rhizophora mangle</i> et <i>Avicennia germinans</i>
	Densification in vegetation cover
	Stability in vegetation cover

**Table 8.** Transition matrix for the mangrove zonation class in Casamance.

Casamance	No Mangrove	High Mangrove	Low and Dense Mangrove	Low and Open Mangrove	Total
No mangrove		70 ha	436 ha	4611 ha	5118 ha
High mangrove	984 ha	4659 ha	4336 ha	2262 ha	12,242 ha
Low and dense mangrove	4284 ha	3263 ha	15,574 ha	12,370 ha	35,491 ha
Low and open mangrove	11,464 ha	124 ha	3472 ha	17,364 ha	32,424 ha
Total	16,732 ha	8116 ha	23,819 ha	36,607 ha	




Legend	
	Decrease spatial extension involving high mangrove dominated by <i>Rhizophora racemosa</i>
	Decrease spatial extension involving low and dense mangrove dominated by <i>Rhizophora mangle</i>
	Decrease spatial extension involving low, open and mixed mangrove dominated by <i>Rhizophora mangle</i> et <i>Avicennia germinans</i>

Table 8. Cont.

Casamance	No Mangrove	High Mangrove	Low and Dense Mangrove	Low and Open Mangrove	Total
					Opening in vegetation cover
					Increase in spatial extension involving high mangrove dominated by <i>Rhizophora racemosa</i>
					Increase spatial extension involving low and dense mangrove dominated by <i>Rhizophora mangle</i>
					Increase spatial extension involving low, open and mixed mangrove dominated by <i>Rhizophora mangle</i> et <i>Avicennia germinans</i>
					Densification in vegetation cover
					Stability in vegetation cover

### 3.3. Spatial Analysis of Changes

#### 3.3.1. Summary Maps of Regeneration

The quantitative analysis shows contrasting dynamics between Saloum and Casamance. First, the colonisation of the high mangrove is uniform in Casamance, whereas it only concentrated in the north-western part of Saloum (Supplementary Materials). This colonisation seems to be taking place on the mudflats available at the edge of the channels. Lower Casamance seems to have available mudflats with moderate salinity throughout its estuary. Saloum appears to have newly viable mudflats only in the region that has experienced marine water intrusion and sedimentary inputs related to the break-up of the Sangomar spit.

The spatial expansion of the low and open mangrove has had an increasingly important contribution to a south–north gradient in Saloum and to a downstream–upstream gradient for Casamance (Figure 8). The low and open mangrove constitutes an indicator of mangrove fronts at the interface with the salt flats, because it has a low degree of canopy closure, which may be indicative of a young mangrove or one at the limits of its habitat. North of Saloum appears to be a suitable place for mangrove colonisation. The mangroves present in the south of Sine–Saloum are becoming denser and are colonising very few new mudflats. Low and open mangrove often occupies the largest proportion of the plant formations contributing to the spatial expansion of the mangrove in the upstream areas of the estuary.

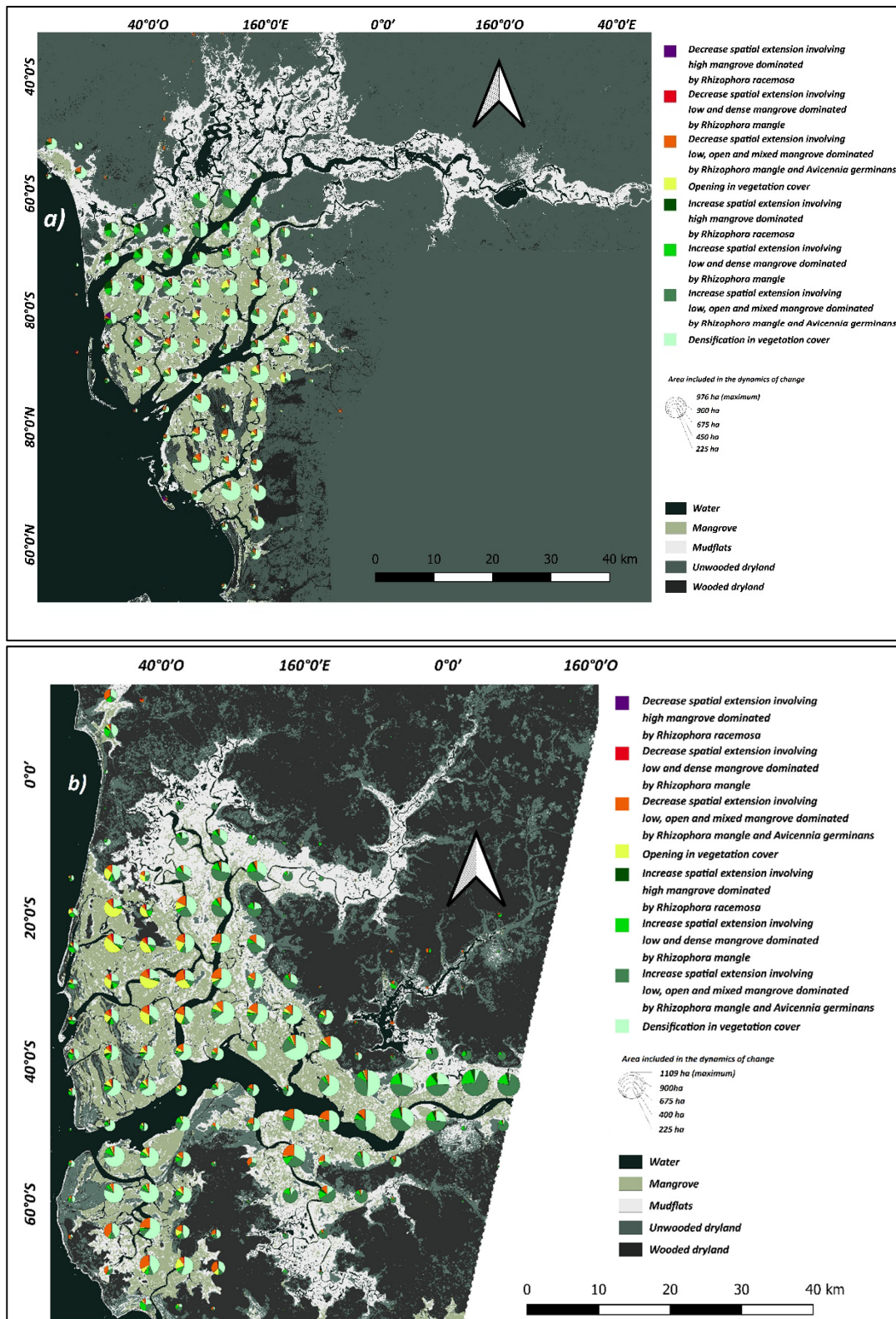
Finally, the densification of the mangrove, i.e., the transition from low and open mangrove to low, dense mangrove, intensifies along an upstream–downstream gradient in the Casamance estuary, unlike in Saloum, where densification has a uniform distribution and substantially contributes to the resilience of the mangrove.

#### 3.3.2. Path Dependence

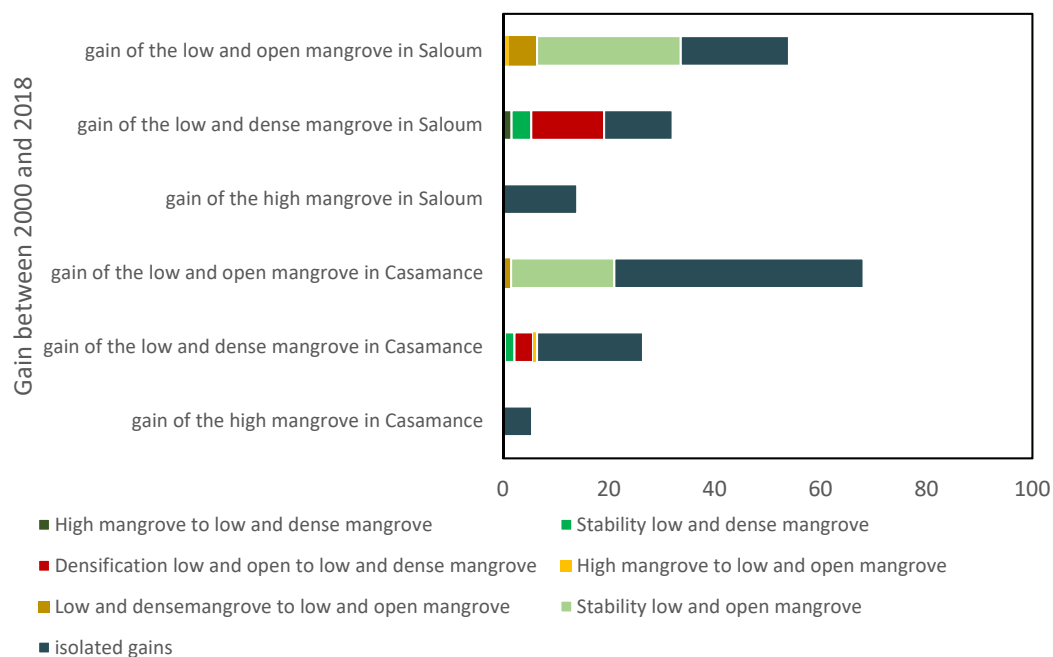
In this paper, the spatiotemporal dynamics of the mangrove was assessed by a transition matrix but, also, by spatial analysis and mathematical morphology. In addition to quantitative changes, treatments sought to measure the extent to which the previous spatial structure of the mangrove conditioned the dynamics of plant formations: the path dependency effect. On the one hand, we measured the percentage of spatial extension of plant formations in contiguity with already present plant formations. On the other hand, we also analysed and quantified through geodetic reconstructions the proportion of each gain in contiguity with pixels where reorganisations in the mangrove were detected. The assumption is made here that the dynamics at work in stable mangrove areas may eventually condition the spatial extension of the mangrove.

In the majority of cases, the new increase of plant formations is not contiguous with pre-existing plant formations (Figure 9), except for the low and open mangrove in Saloum, which grows essentially in contact with its stable nuclei. The proportion of isolated gains in Casamance (72.5%) is much greater than that in Saloum (47%). Densification has played an important role in the expansion of low and dense mangrove in Saloum (13.75%), whereas it is not the main driving force behind the increase in this same class in Casamance (3.5%). The stable nuclei of the low and open mangrove class seem to significantly constrain the spatial distribution of this class in both Saloum (27.2%) and

Casamance (19.50%). The increase in the mangrove area in the latter is therefore mainly conditioned by isolated colonisation or by the propagation of the low and open mangrove front. Saloum is distinguished by similar proportions of isolated and contiguous new gains.



**Figure 8.** Synthesis of the main dynamics of mangrove surfaces in the change process: (a) Saloum and (b) Casamance (the relative sizes of the symbols indicate the importance of the mangrove areas included in the local dynamics).



**Figure 9.** Contribution of the dynamics to the contiguous diffusion of the mangrove areas (the proportions and contributions were evaluated and calculated relative to the total surface areas of the gains, respectively, in the Saloum and Casamance mangroves).

## 4. Discussion

### 4.1. Accuracy and Uncertainty Margin

The previous applications of the same method of stacked unsupervised classification using Landsat images resulted in Kappa indices of more than 0.95 for binary mangrove/non-mangrove mapping for Casamance and Saloum [19]. The application of OBIA using merged Landsat images for plant formations mapping resulted in Kappa indices of 0.83 and 0.80 for Casamance and Saloum. These values are satisfactory compared to the results reported in the literature. Indeed, maps of mangrove vegetation formations obtained by supervised or unsupervised pixel classification with images with similar or higher spatial resolutions had lower Kappa indices [37,46]. According to reference [46], OBIA also perform well (Kappa = 0.83) with RapidEye images (6.5 m/RGB + Rededge + Near infrared) and Pleiades MS images (2 m/RGB + Near infrared). OBIA approaches with Landsat images previously produced Kappa indices ranging from 0.54 to 0.67 [47]. In a study using Sentinel-2 images, Lombard et al. [40] obtained Kappa indices around 0.9 to map plant formations over the same region. The Sentinel-2 images seemed to reduce the margin of error by almost 10% compared to the pansharpened Landsat images. Therefore, the Sentinel-2 satellite could be of great help in the future in monitoring mangroves to guide management policies in mangrove countries.

Based on the contrasts that it generates, the vegetation fraction index appears to increase the ability to discriminate vegetation formations in merged Landsat images. Nevertheless, there is a 20% margin of error that requires the qualification of certain classification results. In Saloum, the confusion between high mangrove dominated by *Rhizophora racemosa* and low, dense mangrove dominated by *Rhizophora mangle* accounts for 19% of the commission errors for the high mangrove class dominated by *Rhizophora racemosa* and 27% of the omission errors for the low, dense mangrove class dominated by *Rhizophora mangle*. Therefore, the densification of *Rhizophora mangle*, which appears to be the main driver of change in the Saloum, may have been underestimated. However, the misclassification rates remain relatively low for the most dynamic plant formations within the two mangrove regions of Senegal. Indeed, the frequencies of confusion between the low



and dense mangrove dominated by *Rhizophora mangle* and the low and open mangrove remain in the acceptable range. The major finding of this study is the confirmation of mangrove dynamics in upper tidal zone. On the other hand, the results do not eliminate uncertainties linked to the discrimination of the canopy height, which is the basis of the separation of the high mangrove dominated by *Rhizophora racemosa* and the low and dense mangrove dominated by *Rhizophora mangle*, particularly in the west of the two regions. However, conclusions about the transitions of these two plant formations are not included in this paper. A way to ensure even more robustness and reduce such inaccuracies could have consisted of a time series analysis, but such method differs significantly from the one used here.

#### 4.2. Summary of Trajectories in the Mangrove Regeneration Process: Densification in the Saloum and Colonisation in Casamance

The mangroves in Senegal experienced undeniable spatial expansion between 2000 and 2018. These results confirm previous findings for both Saloum [20–22,54,55] and Casamance [22,30,56,57]. The method implemented in this work was able to quantify changes in mangrove area at a finer level of detail. This mapping revealed contrasting trajectories between Saloum and Casamance.

Vegetation formations do not equally contribute to the increase in surface area in the two hydrosystems: a densification process is mainly observed in Saloum, whereas colonisation plays an important role in Casamance. First, the low and dense mangrove class is characterised by quasi-monospecific coverage by *Rhizophora mangle*. This plant formation has a high density of individuals and a dense and closed canopy. Conversely, the low and open mangrove class is characterised by a lower density of individuals. Consequently, the transitions observed in this study point to an increase in the density of individuals in Saloum, since the pixels detected as low and open mangrove in 2000 were converted into low and dense mangrove in 2018. This can also be observed in Casamance but to a lesser extent. The reproduction and colonisation of *Rhizophora mangle* thus appear to be important. The proportion of plant formation in mangrove areas has reversed in Saloum, demonstrating the vitality of *Rhizophora mangle*, which now occupies most of the delta. This observation is consistent with population dynamics analyses [20]. In Casamance, a densification process is also underway. However, unlike in Saloum, the low and open mangrove remains stable in these proportions, thus demonstrating an active colonisation process.

In Casamance, the contributions of the changes in densification and mangrove increases are almost identical. On the other hand, the contribution of densification to the observed changes is three times higher in Saloum than in Casamance. Therefore, it appears that the conditions in Saloum are favourable to the reproduction of mangroves, since they are able to densify and grow on mudflats. Nevertheless, these much lower proportions of new gain in Saloum raise questions about the degree to which different species contribute to the dynamics of the low and open mangrove class. Indeed, a previous study [34] identified several gaps between sites in the regeneration of *Avicennia germinans* in Saloum. The only trace of regeneration of the species was stump sprouting. The remote sensing results presented here corroborate these observations on a regional scale. Although *Rhizophora mangle* is the only species to participate in the progression dynamics in Saloum, it is less active than in Casamance, where two species participate in the surface expansion. Moreover, a large proportion of the increase of low and dense mangrove in Saloum (13.75%) is in contact with the densification nuclei that already occupy about 20% of the mangrove surface area. In Saloum, *Rhizophora mangle* occupied at least twice as many viable mudflats as *Avicennia germinans* between 2000 and 2018.

In sum, the results suggest that, in the absence of *Avicennia germinans* reproduction in Saloum, the low and open mangrove class is experiencing a progressive densification of *Rhizophora mangle*. This induces a positive feedback loop in which this dense monospecific stand also drives the monospecific colonisation of the mudflats in the absence of sufficient numbers of *Avicennia germinans*. Furthermore, the significant presence of *Avicennia germinans*

*nans* stump sprouting in Saloum [20,34] could explain the significant stability of the low and open mangrove in this region. The mapping of these dynamics supports previous observations at the local and regional scales [20,34]. Consequently, the plant formations mapped in this study can be regarded as the spatial representation of the biological processes at work in the ecosystem. Efforts must now be directed towards understanding these resilience factors, which lead to contrasting trajectories between Saloum and Casamance.

In summary, two differences can be observed in the dynamics that participate in the spatial expansion of the mangroves in Senegal. The first is the proportion of gain of the low and dense mangrove class and, thus, of *Rhizophora mangle*. In Saloum, the species colonises the mudflats near places where they become dense. In Casamance, the species spreads to places where mangroves were not necessarily present before or grows sporadically, leaving some small discontinuities in the plant cover. The second difference is observed in the respective proportions of isolated new increase in the regions. It is much higher in Casamance than in Saloum.

#### 4.3. Factors in the Dynamics of Plant Formations

The proportion of isolated gain reflects the opportunism of the mangrove or the presence of assisted regeneration. Geodetic reconstructions do not tolerate any pixel discontinuity. Therefore, a gap of 15 m between an existing plant formation and a colonisation causes this increase of surface to be classified as isolated. However, the expansion of mangroves includes not only barochores but, also, hydrochores [58]. The tides can carry propagules to higher positions within tidal zone mudflats, considerably distancing them from the individuals from which they originate [58]. This can explain new gain that is not contiguous with pre-existing plant formations, resulting in the observed discontinuities. These isolated areas of gain can also be an indication of the effects of reforestation programmes. Indeed, in 2000, these practices were common on mudflats devoid of mangroves or in the absence of plant formation [21,59]. In all cases, this progress shows that these areas present viable ecological conditions for the species. Moreover, the colonisation of mudflats occurs by different mechanisms between the two hydrosystems. In Saloum, the mangrove–salt flats interface is the preferred place for the development of mangroves. The absence of fluvial input could explain the low occurrence of accretion zones in the Saloum delta [60–62]. Therefore, mangrove area gains can only be made on the available mudflats located precisely at the mangrove–salt flats interface in Saloum. In Casamance, a non-negligible part of the new gain can be observed on mudflats detected as water on one of the two dates. This corroborates the hypothesis that, on the one hand, these mudflats are in the lower tidal zone and, on the other hand, that Casamance is more prone to sedimentary accretion [61–63] due to more substantial fluvial input compared to Saloum. The annual mean discharge has been recorded at  $24.07 \text{ m}^3 \text{ s}^{-1}$  for Saloum against  $71.4 \text{ m}^3 \text{ s}^{-1}$  for Casamance [23].

The great drought period in West Africa caused an increase in salinities, particularly in the mangrove hydrosystems of Senegal [23,63,64]. A lowering of salinities has been observed in recent years in Saloum and Casamance [23]. Nevertheless, higher salinity values have been measured during the rainy season in Saloum in the absence of significant river input [23]. A strong statistical relationship between salinity levels and the deposition capacity of *Avicennia germinans* propagules in Saloum has been reported [34]. The period of drought [24] led to an opening of the plant cover in the mangrove by the death of certain individuals. On the other hand, the rainfall recovery led to a recolonisation of the mudflats [20] or a filling in of the interstitial spaces, thus causing both a densification of the plant cover and a spatial extension of mangroves. The decrease in the proportions of the low and open mangrove class dominated by *Rhizophora mangle* and *Avicennia germinans* in Saloum seems to coincide with the lack of *Avicennia germinans* reproduction. Moreover, the densification observed by remote sensing supports the hypothesis of a monospecific stand dynamic for *Rhizophora mangle*, which is apparently able to reproduce as suggested [20] in Saloum by a botanical study of progression sites identified by remote sensing. The lack

of *Avicennia germinans* regeneration allows a more competitive species to colonise environments with high salinity levels, which are lower than those during the drought period but higher than those in Casamance. The colonisation observed in Casamance by remote sensing is in line with field observations that showed a balanced age pyramid. It would therefore seem that the lower salinity levels in Casamance, since the rainfall recovery in the mid-1990s has enabled the reproduction of *Avicennia germinans* and *Rhizophora mangle*, which could explain the colonisation mapped in this study. In Saloum, our results show the emergence of a different regional dynamic that prompts further analysis of the biological processes in the delta. Furthermore, the significant densification of *Rhizophora mangle*, as well as its considerably restricted spatial expansion in Saloum compared to that in Casamance, supports the hypothesis of an absence of satisfactory environmental conditions for the reproduction of *Avicennia germinans* on a regional scale [20,34]. It is therefore necessary to gain a more detailed understanding of the links between fluctuating environmental conditions and the success (or lack thereof) of the seedling processes that underlie the population dynamics of mangrove species.

## 5. Conclusions

The objective of this study was to monitor mangrove plant formations in Senegal between 2000 and 2018. Using an OBIA approach with images of the vegetation fraction resulting from linear demixing, a more detailed quantification of the dynamics at work in Senegalese mangroves was carried out. These mangroves have experienced an increase in their surface area. Nevertheless, the trajectory of Senegalese mangroves differs between areas. The results suggest that the different species making up these formations have disparate successes of reproduction and diffusion. A clear increase in low and dense mangrove and substantial changes from low and open mangrove to low and dense mangrove demonstrate that the mangrove densification of *Rhizophora mangle* is due, in part, to the failure of *Avicennia germinans* to reproduce. This process shows that, despite the absence of *Avicennia germinans* reproduction, the Saloum mangrove exhibits significant resilience, which, moreover, does not seem to benefit from significant regeneration, which had an estimated contribution of 4% [20]. However, the frequency of stable cores of low and open mangrove remains low in the northern three-quarters of Saloum. The hypothesis verified in this work and emanating from field observations [20,34] is corroborated by this remote sensing study at the regional scale. The biological processes at the local scale seem to cause a reorganisation of the mangrove at the regional scale. This suggests that gaps in *Avicennia germinans* regeneration may eventually prevent the mangrove from continuing its progression on the upper tidal zone in Saloum. The regeneration gaps of *Avicennia germinans* impact the expansion of the Saloum mangrove, which raises the question of the relevance of reforestation policies that are based mainly on the genus *Rhizophora* spp. and that are mainly undertaken in Casamance. It would therefore seem appropriate to reassess conservation policies in Senegalese mangroves. First, reforestation in Saloum does not seem to be effective and should target *Avicennia germinans* as a priority. Second, as the mangroves are showing resilience and have increased in surface areas, optimising reforestation policies could eventually lead to sustainable forestry in the Senegalese mangrove instead of the current restrictive policy. Although strict conservation and wilderness reserve approaches have proven their worth in many regions, it is still useful to explore other paths to achieving conservation through sustainable management of the resources. The diachronic mapping established in this work could be used to monitor the evolution of plant formations corresponding to landscape elements that are perceived and managed as resources by local societies.

**Supplementary Materials:** The following are available online at <https://www.mdpi.com/article/10.3390/rs13101961/s1>, KML vector files.

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## References

- Friess, D.A.; Rogers, K.; Lovelock, C.E.; Krauss, K.W.; Hamilton, S.E.; Lee, S.Y.; Lucas, R.; Primavera, J.; Rajkaran, A.; Shi, S. The State of the World's Mangrove Forests: Past, Present, and Future. *Annu. Rev. Environ. Resour.* **2019**, *44*, 89–115. [\[CrossRef\]](#)
- Lee, S.; Hamilton, S.; Barbier, E.; Primavera, J.; Lewis, R. Better restoration policies are needed to conserve mangrove ecosystems. *Nat. Ecol. Evol.* **2019**, *3*, 870–872. [\[CrossRef\]](#) [\[PubMed\]](#)
- Otero, V.; Van De Kerchove, R.; Satyanarayana, B.; Mohd-Lokman, H.; Lucas, R.; Dahdouh-Guebas, F. An analysis of the early regeneration of mangrove forests using landsat time series in the Matang mangrove forest reserve, Peninsular Malaysia. *Remote Sens.* **2020**, *11*, 774. [\[CrossRef\]](#)
- Duke, N.C.; Meynecke, J.-O.; Dittmann, S.; Ellison, A.M.; Anger, K.; Berger, U.; Cannicci, S.; Diele, K.; Ewel, K.C.; Field, C.D.; et al. A World Without Mangroves? *Science* **2007**, *317*, 41–42. [\[CrossRef\]](#) [\[PubMed\]](#)
- IPBES. *Global Assessment Report on Biodiversity and Ecosystem Services of the Intergovernmental Science-Policy Platform on Biodiversity and Ecosystem Services*; Brondizio, E.S., Settele, J., Díaz, S., Ngo, H.T., Eds.; IPBES secretariat: Bonn, Germany, 2019.
- Donato, D.C.; Kauffman, J.B.; Murdiyarto, D.; Kurnianto, S.; Stidham, M.; Kanninen, M. Mangroves among the most carbon-rich forests in the tropics. *Nat. Geosci.* **2011**, *4*, 293–297. [\[CrossRef\]](#)
- Alongi, D.M. *Introduction in the Energetics of Mangrove Forests*; Springer Science and Business Media BV: New York, NY, USA, 2012.
- Friess, D.A.; Yando, E.S.; Abuchahla, G.M.O.; Adams, J.B.; Cannicci, S.; Canty, S.W.J.; Wee, A.K.S. Mangroves give cause for conservation optimism, for now. *Curr. Biol.* **2020**, *30*, R153–R154. [\[CrossRef\]](#)
- Saenger, P. *Mangrove Ecology, Silviculture and Conservation*; Kluwer Academic Publishers: Dordrecht, Netherlands, 2002; pp. 11–18.
- Spalding, M.; Kainuma, M.; Collins, L. *World Atlas of Mangroves*; A Collaborative Project of ITTO, ISME, FAO, UNEP-WCMC, UNESCO-MAB, UNU-INWEH and TNC. 2010; Earthscan: London, UK, 2010; p. 319.
- Gallup, L.; Sonnenfeld, D.A.; Dahdouh-Guebas, F. Mangrove use and management within the Sine-Saloum Delta, Senegal. *Ocean Coast. Manag.* **2020**, *185*, 105001. [\[CrossRef\]](#)
- Walters, B.B.; Rönnbäck, P.; Kovacs, J.M.; Crona, B.; Hussain, S.A.; Badola, R.; Primavera, J.H.; Barbier, E.; Dahdouh-Guebas, F. Ethnobiology, socio-economics and management of mangrove forests: A review. *Aquat. Bot.* **2008**, *89*, 220–236. [\[CrossRef\]](#)
- Cormier-Salem, M. *Dynamique et Usages de la Mangrove Dans les Pays des Rivières du Sud, du Sénégal à la Sierra Leone*; IRD Éditions: Marseille, France, 1991. [\[CrossRef\]](#)
- Dahdouh-Guebas, F.; Collin, S.; Seen, D.L.; Rönnbäck, P.; Depommier, D.; Ravishankar, T.; Koedam, N. Analysing ethnobotanical and fishery-related importance of mangroves of East-Godavari Delta (Andhra Pradesh, India) for conservation and management purposes. *J. Ethnobiol. Ethnomed.* **2006**, *2*, 24. [\[CrossRef\]](#)
- Edwards, P. Aquaculture environment interactions: Past, present and likely future trends. *Aquaculture* **2015**, *447*, 2–14. [\[CrossRef\]](#)
- FAO. *The World's Mangroves 1980–2005*; Étude FAO: Rome, Italy, 2007.
- Bryan-Brown, D.N.; Connolly, R.M.; Richards, D.R. Global trends in mangrove forest fragmentation. *Sci. Rep.* **2020**, *10*, 7117. [\[CrossRef\]](#) [\[PubMed\]](#)
- FAO. *Global Forest Resources Assessment 2020: Main Report*; FAO: Rome, Italy, 2020. [\[CrossRef\]](#)
- Andrieu, J. Land cover changes on the West-African coastline from the Saloum Delta (Senegal) to Rio Geba (Guinea-Bissau) between 1979 and 2015. *Eur. J. Remote Sens.* **2018**, *51*, 314–325. [\[CrossRef\]](#)
- Andrieu, J.; Lombard, F.; Fall, A.; Thior, M.; Demba Ba, B.; Ephrem, A.; Dieme, B. Botanical field-study and remote sensing to describe mangrove resilience in the Saloum Delta (Senegal) after 30 years of degradation narrative. *For. Ecol. Manag.* **2020**, *461*, 117963. [\[CrossRef\]](#)
- Dieye, E.H.B.; Diaw, A.T.; Sané, T.; Ndour, N.; Ndour, N. Dynamique de la mangrove de l'estuaire du Saloum (Sénégal) entre 1972 et 2010. *Cybergeogr. Eur. J. Geogr.* **2013**, doc 629. [\[CrossRef\]](#)



22. Fent, A.; Bardou, R.; Carney, J.; Cavanaugh, K. Transborder political ecology of mangroves in Senegal and The Gambia. *Glob. Environ. Chang.* **2019**, *54*, 214–226. [[CrossRef](#)]
23. Descroix, L.; Sané, Y.; Thior, M.; Manga, S.-P.; Ba, B.D.; Mingou, J.; Mendy, V.; Coly, S.; Dièye, A.; Badiane, A.; et al. Inverse Estuaries in West Africa: Evidence of the Rainfall Recovery? *Water* **2020**, *12*, 647. [[CrossRef](#)]
24. Nicholson, S.E. The West African Sahel: A review of recent studies on the rainfall regime and its interannual variability (Article ID 453521). *ISRN Meteorol.* **2013**, *32*. [[CrossRef](#)]
25. Mahé, G.; Lienou, G.; Descroix, L.; Bamba, L.; Paturel, J.-E.; Laraque, A.; Meddi, M.; Habaieb, M.; Adeaga, O.; Dieulin, C.; et al. The rivers of Africa: Witness of climate change and human impact on the environment. *Hydrol. Process.* **2013**, *27*, 2105–2114. [[CrossRef](#)]
26. Moreno-de las Heras, M.; Díaz-Sierra, R.; Turnbull, L.; Wainwright, J. Assessing vegetation structure and ANPP dynamics in a grassland–shrubland Chihuahuan ecotone using NDVI–rainfall relationships. *Biogeosciences* **2015**, *12*, 2907–2925. [[CrossRef](#)]
27. Thoma, D.P.; Munson, S.M.; Irvine, K.M.; Witwicki, D.L.; Bunting, E.L. Semi-arid vegetation response to antecedent climate and water balance windows. *Appl. Veg. Sci.* **2016**, *19*, 413–429. [[CrossRef](#)]
28. Zhou, J.; Jia, L.; Menenti, M.; Van Hoek, M.; Lu, J.; Zheng, C.; Wu, H.; Yuan, X. Characterizing vegetation response to rainfall at multiple temporal scales in the Sahel-Sudano-Guinean region using transfer function analysis. *Remote Sens. Environ.* **2021**, *252*, 112108. [[CrossRef](#)]
29. Holling, C.S. *Adaptive Environmental Assessment and Management*; John Wiley and Sons: Chichester, UK, 1978.
30. Conchedda, G.; Durieux, L.; Mayaux, P. An object-based method for mapping and change analysis in mangrove ecosystems. *ISPRS J. Photogramm. Remote Sens.* **2008**, *63*, 578–589. [[CrossRef](#)]
31. Cormier-Salem, M.C. *Mangrove Grabbing: An Exploration of Changes in Mangrove Tenure from a Political Ecology Perspective*; The Sea within: Maritime Tenure and Cosmopolitical Debates; Artaud, H., Surallés, A., Eds.; IWGIA: Copenhagen, Denmark, 2017; pp. 143–162.
32. Cormier-Salem, M.-C.; Panfili, J. Mangrove reforestation in question: Greening or grabbing coastal zones and deltas? *Afr. J. Aquat. Sci.* **2016**, *41*, 89–98. [[CrossRef](#)]
33. Arumugam, M.; Niyomugabo, R.; Dahdouh-Guebas, F.; Hugé, J. The perceptions of stakeholders on current management of mangroves in the Sine-Saloum Delta, Senegal, Estuarine. *Coast. Shelf Sci.* **2021**, *248*, 107160. [[CrossRef](#)]
34. Lombard, F.; Andrieu, J.; Descroix, L. La population d’*Avicennia germinans* du delta du Saloum est-elle relictuelle depuis la dernière période humide? *Bois et Forêts des Trop.* **2020**, *346*, 51–64. [[CrossRef](#)]
35. Giri, C.; Ochieng, E.; Tieszen, L.L.; Zhu, Z.; Singh, A.; Loveland, T.; Masek, J.; Duke, N. Status and distribution of mangrove forests of the world using earth observation satellite data. *Glob. Ecol. Biogeogr.* **2011**, *20*, 154–159. [[CrossRef](#)]
36. Kuenzer, C.; Bluemel, A.; Gebhardt, S.; Quoc, T.V.; Dech, S. Remote Sensing of Mangrove Ecosystems: A Review. *Remote Sens.* **2011**, *3*, 878–928. [[CrossRef](#)]
37. Wang, D.; Wan, B.; Qiu, P.; Su, Y.; Guo, Q.; Wang, R.; Sun, F.; Wu, X. Evaluating the Performance of Sentinel-2, Landsat 8 and Pléiades-1 in Mapping Mangrove Extent and Species. *Remote Sens.* **2018**, *10*, 1468. [[CrossRef](#)]
38. Taureau, F.; Robin, M.; Proisy, C.; Fromard, F.; Imbert, D.; Debaine, F. Mapping the Mangrove Forest Canopy Using Spectral Unmixing of Very High Spatial Resolution Satellite Images. *Remote Sens.* **2019**, *11*, 367. [[CrossRef](#)]
39. Wang, L.; Sousa, W.P.; Gong, P. Integration of object-based and pixel-based classification for mapping mangroves with IKONOS imagery. *Int. J. Remote Sens.* **2004**, *25*, 5655–5668. [[CrossRef](#)]
40. Lombard, F.; Andrieu, J.; Josselin, D. Object-oriented classification-based mapping of mangrove zonation in Casamance and Sine-Saloum (Senegal) using Sentinel-2 images. *Geo-Spat. Inf. Sci.* **2021**. submitted for publication.
41. Johnson, B. Effects of Pansharpening on Vegetation Indices. *ISPRS Int. J. Geo-Inf.* **2014**, *3*, 507–522. [[CrossRef](#)]
42. Kasawani, I.; Kim, A.; Shah, A.; Sulong, I.; Ismail, M.H.; Jusoff, K. Coastal Change Detection using GIS in Setiu Lagoon, Terengganu, Malaysia. *J. GIS Trends* **2010**, *1*, 20–26.
43. Yuvaraj, E.; Dharanirajan, K.; Saravanan, N.; Karpoorasundarapandian, N. Evaluation of Vegetation density of the Mangrove forest in South Andaman Island using Remote Sensing and GIS techniques. *Int. Res. J. Environ. Sci.* **2014**, *3*, 19–25.
44. Sari, S.P.; Rosalina, D. Mapping and Monitoring of Mangrove Density Changes on tin Mining Area. *Procedia Environ. Sci.* **2016**, *33*, 436–442. [[CrossRef](#)]
45. Myint, S.; Chandra, G.; Wang, L.; Gillette, S. Identifying Mangrove Species and Their Surrounding Land Use and Land Cover Classes Using an Object-Oriented Approach with a Lacunarity Spatial Measure. *GISci. Remote Sens.* **2008**, *45*, 188–208. [[CrossRef](#)]
46. Taureau, F. *Cartographie Harmonisée des Mangroves de L’outre-Mer Français*. Thèse de Doctorat. Sciences de L’environnement; Université de Nantes (UNAM), Nantes, France, 2017. Doctorat Thèse, Sciences de L’environnement. Université de Nantes (UNAM), Nantes, France, 2017.
47. Bullock, E.L.; Fagherazzi, S.; Nardin, W.; Vo-Luong, P.; Nguyen, P.; Woodcock, C.E. Temporal patterns in species zonation in a mangrove forest in the Mekong Delta, Vietnam, using a time series of Landsat imagery. *Cont. Shelf Res.* **2017**, *147*, 144–154. [[CrossRef](#)]
48. Gudex-Cross, D.; Pontius, J.; Adams, A. Enhanced forest cover mapping using spectral unmixing and object-based classification of multi-temporal Landsat imagery. *Remote Sens. Environ.* **2017**, *196*, 193–204. [[CrossRef](#)]
49. Descroix, L. *Processus et Enjeux D’eau en Afrique de L’ouest Soudano-Sahélienne*; EAC: Paris, France, 2018; p. 320. ISBN 978-2-8130-0314-0.



50. Andrieu, J.; Cormier-Salem, M.C.; Descroix, L.; Diéye, E.H.B.; Ndour, N.; Sané, T. Correctly assessing forest change in a priority West African mangrove ecosystem: 1986–2010 An answer to Carney et al., 2014 paper “Assessing forest change in a priority West African mangrove ecosystem: 1986–2010”. *Remote Sens. App. Soc. Environ.* **2019**, *13*, 337–347. [[CrossRef](#)]
51. Datta, D.; Roy, A.K.; Kundu, A.; Dutta, D.; Neogy, S. An alternative approach to delineate wetland influence zone of a tropical intertidal mudflat using geo-information technology. *Estuar. Coast. Shelf Sci.* **2021**, *253*, 107308. [[CrossRef](#)]
52. Taureau, F. *Evaluation des Données Sentinel Pour la Mise en Place D'un Système de Surveillance des Mangroves à Mayotte*; Université de Nantes: Nantes, France, 2018; p. 116.
53. Landis, J.R.; Koch, G.G. The measurement of observer agreement for categorical data. *Biometrics* **1977**, *33*, 159–174. [[CrossRef](#)]
54. Olofsson, P.; Foody, G.M.; Herold, M.; Stehman, S.V.; Woodcock, C.E.; Wulder, M.A. Good practices for estimating area an assessing accuracy of land change. *Remote Sens. Environ.* **2014**, *148*, 42–57. [[CrossRef](#)]
55. Sakho, I.; Mesnage, V.; Deloffre, J.; Lafite, R.; Niang, I.; Faye, G. The influence of natural and anthropogenic factors on mangrove dynamics over 60 years: The Somone Estuary Senegal. *Estuar. Coast. Shelf Sci.* **2011**, *94*, 93–101. [[CrossRef](#)]
56. Solly, B.; Diéye, E.H.B.; Sané, T.; Diaw, A.T. Dynamique de la Mangrove de Thiobon dans l'estuaire de la Casamance (Sénégal) entre 1972 et 2017. *Eur. Sci. J.* **2018**, *14*, 118–133. [[CrossRef](#)]
57. Soumaré, S.; Fall, F.; Andrieu, J.; Marega, O.; Dieme, B. Dynamique spatio-temporelle de la mangrove de Kafountine dans l'estuaire de la Basse-Casamance des années 1972 à nos jours: Approche par télédétection. *IOSR J. Eng. (IOSRJEN)* **2020**, *10*, 1–14.
58. Hogarth, P.J. *The Biology of Mangroves and Seagrasses*; Biology of Habitats, 2 ed.; Oxford University Press: Oxford, UK, 2007; ISBN 978-0-19-856870-4.
59. Bassene, O.; Cubizolle, H.; Cormier-Salem, M.C.; Sy, B.A. L'impact des changements démographiques et socio-économiques sur la perception et la gestion de la mangrove en Basse Casamance (Sénégal). *Géocarrefour* **2013**, *88*, 299–315. [[CrossRef](#)]
60. Barusseau, J.-P.; Diop, E.S.; Saos, J.-L. Evidence of dynamics reversal in tropical estuaries, geomorphological and sedimentological consequences (Salum and Casamance Rivers, Senegal). *Sedimentology* **1985**, *32*, 543–552. [[CrossRef](#)]
61. Diop, E.S. La Côte Ouest Africaine du Saloum (Sénégal) à la Mellacorie (République de Guinée). Ph.D. Thesis, Université Louis Pasteur, Strasbourg, France, 1990.
62. Marius, C. Mangroves du Sénégal et de la Gambie. Ecologie, Pédologie, Géochimie, Mise en Valeur et Aménagement. Ph.D. Thesis, Louis Pasteur Strasbourg University, Strasbourg, France, 1985; p. 335.
63. Diouf, A.C. Dynamiques Des Ressources Naturelles et Adaptations Des Sociétés Dans la Réserve de Biosphère du Delta du Saloum au Sénégal. Ph.D. Thesis, Université Gaston Bergé, Dakar, Senegal, 2019.
64. Olivry, J.-C. Les conséquences durables de la sécheresse actuelle sur l'écoulement du fleuve Sénégal et l'hypersalinisation de la Basse-Casamance. In *Influence of Climate Change and Climatic Variability on the Hydrologic Regime and Water Resources*; IAHS Publisher: Wallingford, UK, 1987.