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Radarsat-2 Backscattering for the Modeling of Biophysical Parameters of Regenerating Mangrove Forests

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Abstract: The aim of this study is to understand the relationship between radar backscattering (σ° , β° and γ) of a multi-polarized Radarsat-2 C-band image with the structural attributes of regenerating mangrove vegetation located at the mouth of the Amazon River. CBH (circumference at breast height), height and species data were collected to characterize vegetation structure and above-ground biomass (AGB) at 17 plots with a total of 3090 measured individuals. Significant relationships between the linear σ° in VH (vertical transmit, horizontal receive) cross-polarization produced r^2 values of 0.63 for the average height, 0.53 for the DBH, 0.46 for the basal area (BA) and 0.52 for the AGB. Using co-polarized HH (horizontal transmit, horizontal receive) and VV (vertical transmit, vertical receive), r^2 values increased to 0.81, 0.79, 0.67 and 0.79, respectively. Vegetation attribute maps of average canopy height, DBH and AGB were generated for the study area. We conclude that multi-polarized Radarsat-2 images were adequate for characterization of vegetation attributes in areas of mangrove regeneration.

Keywords: SAR; regenerating mangrove; biomass; Amazon

1. Introduction

Mangroves are among the most productive ecosystems in the world [1], exporting nutrients and organic matter to adjacent waters and coastal environments, and providing for a complex aquatic food web [1,2]. Mangroves have great economic and ecological significance, protecting and stabilizing the coastline, and acting as nurseries and breeding grounds for numerous wildlife species, valuable goods and services [3–6].

Mangrove productivity is directly linked to biomass, which is important for understanding the cycling of organic matter in mangrove ecosystems [6]. A traditional method of estimating biomass is by manually measuring structural parameters of vegetation through allometric equations. Measuring these parameters by non-destructive means is a challenge that has been reported by many authors in Africa [7], Europe [8], the Americas, Asia, and Oceania [9–13].

Research studies have attempted to produce inventories and establish efficient approaches for the monitoring and conservation of mangroves [1,3,5,14,15]. However, these ecosystems are difficult to access because of the maze of roots and stems, and unconsolidated substrate and flood tides [16]. Thus, remote sensing imaging with different spectral frequencies and spatial and temporal resolutions has proven to be a more efficient source of data to study the dynamics of mangrove forests at large scales [17–22]. This paper investigates the use of synthetic aperture radar (SAR) in regenerating mangrove forests. Radar instruments operate in the microwave spectrum and make a suitable sensor to monitor low-latitude environments characterized by the frequent presence of clouds, rain and smoke.

Radar backscatter results from microwaves reflected from vegetation components, such as twigs, branches and trunks [23,24]. Several studies have related radar backscattering with the structural parameters of mangrove vegetation such as homogeneous forest canopies to estimate above-ground biomass (AGB) [25–29]. Recently, Kovacs *et al.* [30,31] estimated structural attributes of degraded mangrove forests on the Pacific coast of México using multi-polarized C-band (Radarsat-2) and L-band images (ALOS PALSAR). The mangrove area is the focus of this study and it is located in the Bragança Peninsula (northeastern coast of Pará) along the northern coast of Brazil. The region has undergone significant anthropic pressure in the last 30 years because of the construction of a highway to facilitate access to coastal resources by the local population and allow mangrove products to be transported to local markets [32]. The highway slices the intertidal mud flat deposits that are densely colonized by mangrove forests over a stretch of 25 km, thus modifying the hydrological regime and causing significant die-off of vegetation that was subsequently removed by the local population. After a few years, part of this degraded area showed incipient natural regeneration [21].

This study aims to evaluate the relationship between the structural attributes of the regenerating mangrove vegetation and multi-polarized data from the Radarsat-2 (C-band) sensor using statistical regression models. The correlation between radar attributes (*i.e.*, σ° , β° and γ) at the four polarization configurations (horizontal transmit, horizontal receive—HH, horizontal transmit, vertical receive—HV, vertical transmit, horizontal receive—VH and vertical transmit and vertical receive—VV) and the mangrove structure data (diameter at breast height—DBH, basal area—BA, height and biomass) was investigated. Finally, the regression models are used to generate forest structure maps of a regenerating mangrove in order to support implementation of rehabilitation and restoration efforts.

2. Data and Methods

2.1. Study Area

The study area is located along the northern coast of Brazil approximately 380 km southeast of the mouth of the Amazon River (Figure 1). It is part of the largest continuous mangrove belt in the world [22,33]. The climate is classified as hot and humid [34] with two seasons, rainy (January to July) and dry (August to January), which have an average rainfall of 2000 mm and 20 mm, respectively [35]. The region boasts a semidiurnal macrotidal regime with minimum variations of 1.8 m and maximum variations of 5.4 m [36]. The mangroves of the Bragança Peninsula occupy an area of 466 km² over extensive mud flats up to 30 km wide and are located between the high levels of spring and mean tide level [37].

The floristic composition of the mangrove vegetation in the region is composed of four mangrove species: *Rhizophora mangle* L. (Rhizophoraceae), *Avicennia germinans* (L.) L., *Avicennia schaueriana* Stapf and Leechman (Acanthaceae) and *Laguncularia racemosa* (L.) CF. Gaertn. (Combretaceae). The species *R. mangle* is predominant [38]. Despite the low variety in species, there is a great variability in the structure of the mangrove forest because of topography and local hydrological conditions [39].



Figure 1. Map of the study area: **(A)** SRTM and **(B)** RapidEye image in 1R2G3B normal color composition. The figure also shows the location of the 17 plots analyzed in this study.

2.2. SAR Data and Image Processing

The SAR data used in this investigation were a multi-polarized Radarsat-2 image (Table 1) obtained in fine-beam mode (FQ5). Precipitation and tide data were provided by the National Institute of Meteorology [40] and Directorate of Hydrography and Navigation of the Brazilian Navy [41], respectively. An optical image obtained with the REIS (RapidEye Earth Imaging System) sensor acquired on 18 July 2011 was used to facilitate the location of field plots.

Table 1. Characteristics of the SAR image and environmental conditions on the acquisition day.

Sensor RADARSAT-2								
Frequency GHz (band)	5.40 (C)							
Wavelength	5.6 cm							
Polarization	HH/HV/VH/VV							
Acquisition mode	Fine Quad-Pol							
Level processing	Single Look Complex–SLC							
Data type (n looks)	Polarimetric (1)							
Nominal resolution	$11 \times 9 \text{ m}$							
Pixel spacing	4.73 (range) and 4.98 (azimuth)							
Orbit of acquisition	Descending							
Acquisition date	6 November 2010							
Time of acquisition	08:55:58							
Incidence angle	23.39°-25.28°							
Precipitation	no rain							
Tide Condition	+3 m							

The Shuttle Radar Topography Mission (SRTM) digital elevation model with 11 ground control points (GCP) was used for image orthorectification. This step was developed with the use of Toutin's Radargrammetic model [42] implemented in the application OrthoEngine/PCI; the Root Mean Square Error (RMSE) was 12.2 m, 12.19 m and 12.58 m for the σ° , β° and γ images, respectively. Subsequently, the speckle noise was reduced with the use of the enhanced Lee filter [43]. Among the three applied window sizes (3 × 3, 5 × 5 and 7 × 7 pixels), the 5 × 5 pixel size was shown in a

previous study to provide a better performance in the analysis of the correlations with the biophysical parameters. These processing steps were performed for the three reflectivity parameters (σ° , β° and γ), representing the reflected microwave as: sigma (σ°) the average reflectivity of a sample of target normalized by the unit area in the ground range; gamma (γ) the reflectivity measured in terms of forward incident wave; and beta (β°) the reflectivity in slant range which is independent of the local incident angle [16]. The use of parameter σ° in studies based on backscattering is a consensus in the literature [16,26]. However, we also investigated the relationship of γ and β° parameters with vegetation parameters based on the fact that γ will remain approximately constant for all incidence angles, and thus is a more convenient measurement parameter to employ than σ° when dealing with volume scattering targets, such as forest. The β° parameter is independent of the local topography and it represents the only directly measure from image radar system, what is known as "radar brightness" [44]. Subsequently, the average backscattering values of the three radar attributes were extracted for each investigated plot of 100 m².

2.3. Collecting Structural Data in the Mangrove Forest

Initially, before fieldwork, six classes were visually distinguished in the RapidEye image to determine the local vegetation gradient: exposed ground, recent stage, initial regeneration, intermediate regeneration, advanced regeneration, and flooded vegetation. To extract backscattering values from the Radarsat-2 image, 20 plots were defined over the SAR image to characterize the six initial classes, totaling 120 plots. To reduce the number of plots to be inventoried during the fieldwork, a plot cluster analysis was performed based on the backscattering values. Hence, 17 plots were selected to be inventoried during the fieldwork carried out in August and December 2012. The central and corner coordinates of each plot were determined with the use of a differential global positioning system (DGPS—model ASTECH Z-Xtreme) with decimeter accuracy. The determination of the corner coordinates of each plot was carried out with a TOPCON total station model GTS 210. The position of each tree and shrub present in the plot were obtained in the same manner. The size of the plots (10×10 m) was defined according to the nominal resolution (11×9 m) of the Radarsat-2 Fine Quad Pol 5 mode, equivalent to approximately four pixels of this SAR image; this plot size is adequate due to high mangrove vegetation density [45].

The botanical species were identified and biophysical parameters, such as the circumference at breast height (CBH) and height of the individuals, were measured in each plot with their respective geographical positions. For low stature trees without trunks at 1.30 m (CBH), the circumference measurement was performed below the first branch, which was proposed by Soares [46]. Subsequently, the diameters at breast height (DBH), basal area (BA), mean and maximum heights and density values were calculated according to Cíntron and Schaeffer Novelli [45]. Lorey's height was also calculated for each plot [47]. After the collection of biophysical parameters, exposed ground class without vegetation was not included in the statistical analysis and flooded vegetation was recognized as the advanced regeneration class. Hence, four classes were redefined as recent stage, initial regeneration, intermediate regeneration and advanced regeneration. The recent stage is characterized as being bare soil with recent colonization of single seedlings of mangrove vegetation (average BA of 2.9 $m^2 \cdot ha^{-1}$). The initial stage of regeneration has less exposed soil due to occurrence of small shrubs (average BA of 9 m² \cdot ha⁻¹). In the intermediate regeneration stage the soil is covered by vegetation with more structural development (average BA of 14 m² \cdot ha⁻¹), but lower than that of the advanced regeneration stage, where the trees can reach 15 m in height and an average BA of $20 \text{ m}^2 \cdot \text{ha}^{-1}$ (Figure 2).

After the biophysical data were processed, a cluster analysis was performed to associate plots with similar structural development of the canopy. The average canopy height, DBH and BA were analyzed through the Euclidean distance method. An analysis of variance (ANOVA) was applied to the formed clusters to investigate significant differences in the distribution of the structural

parameters. Subsequently, the *post-hoc* Tukey's test [48] characterized the differences within the clusters by multiple comparisons of the paired clusters.



Figure 2. (**A**) Cluster analysis using Euclidian distance with the biophysical parameters: average canopy height, DBH and basal area; (**B**) Different mangrove regeneration stages observed in the field.

2.4. Estimation of the Above-Ground Biomass

The allometric equations proposed by Fromard *et al.* [49] for the mangroves of French Guiana were used to estimate the AGB of the study area. There are no specific allometric equations for mangrove trees along the coast of the Brazilian Amazon. When developing Equations (1) to (4), Fromard *et al.* [49] indicated that the independent variable DBH was used because it is a parameter that can be measured for all individuals more accurately than height.

Avicennia germinans :
$$1 \text{ cm} < \text{DBH} < 4 \text{ cm} : \text{y} = 200.4 \text{ DBH}^{2,1}(\text{g})$$
 (1)

$$DBH > 4 \,cm: \, y = 0.14 \, DBH^{2,4} \, (Kg) \tag{2}$$

Laguncularia racemosa : $y = 102.3 \text{ DBH}^{2,5}(g)$ (3)

$$Rhizophora \operatorname{spp.}: y = 128.2 \operatorname{DBH}^{2,6}(g) \tag{4}$$

Based on the above equations, our dataset spans a range of AGB of 0.5–2.8 kg (DBH < 4 cm) and 5.7–1543.7 kg (DBH > 4 cm) for *A. germinans*, 0.3–27.6 kg for *L. racemosa* and 0.3–1036.6 kg for *Rhizophora* spp. Plots were then rearranged in order of increasing AGB to facilitate the presentation of data and discussion.

2.5. Modeling the Impact of Forest Structure in Regenerating Mangrove on SAR Data

Analysis of the relationship between the structural attributes and the multi-polarized backscattering of the Radarsat-2 image was performed using simple and multiple regression statistical methods in which the independent variables were the backscattering values and structural attribute values were the dependent variables. The development of the models followed the methodology described by Neter *et al.* [50] with various functions: linear, logarithmic, second- and third-order polynomial, power and exponential.

In the multiple linear regression model, the selection of the variables was based on the best subset [50] regressive method and decision criteria (r^2 , r^2 fit and Cp Mallow) in which the best fit with the fewest possible explanatory variables is identified. The validation of the developed models was performed by the methods PRESS (Prediction Sum of Squares) and RMSE (Root Mean Square Error).

3. Results and Discussion

3.1. SAR Attributes of the Mangrove Features

The mean backscattering values on a linear scale extracted from the Radarsat-2 image for each of the studied plots are shown in Table 2. The cross-polarization channels showed backscattering values lower than the co-polarization channels for all reflectivity parameters. While the expected strong surface and double-bounce scattering is observed in co-polarized images, lower backscatter is observed at cross-polarizations, which results mainly from the volume scattering occurring within the mangrove canopies [26].

Table 2. Mean backscattering values in σ° , β° and γ extracted for each plot (P). See Figure 3 for the location of the plots.

	Р	${\sigma^\circ}_{HH}$	${\sigma^\circ}_{HV}$	${\sigma^\circ}_{VH}$	${\sigma^\circ}_{VV}$	$\beta^_{HH}$	β°_{HV}	$\beta^\circ v _{H}$	β°_{VV}	γнн	$\gamma_{\rm HV}$	γ_{VH}	γvv
	1	0.117	0.009	0.011	0.242	0.145	0.025	0.022	0.556	0.181	0.011	0.015	0.26
	2	0.131	0.002	0.001	0.065	0.28	0.012	0.01	0.215	0.16	0.017	0.007	0.165
Recent stage	3	0.148	0.029	0.03	0.077	0.415	0.123	0.098	0.299	0.164	0.041	0.041	0.149
Recent stage	4	0.018	0.006	0.006	0.109	0.071	0.019	0.015	0.209	0.069	0.009	0.006	0.173
	5	0.045	0.028	0.039	0.301	0.298	0.057	0.085	0.284	0.189	0.027	0.048	0.133
	6	0.319	0.033	0.03	0.696	1.242	0.101	0.104	2.355	0.082	0.037	0.029	0.522
	7	0.104	0.027	0.023	0.088	0.535	0.092	0.05	0.411	0.175	0.015	0.015	0.087
	8	0.241	0.002	0.005	0.063	0.447	0.02	0.006	0.214	0.352	0.032	0.024	0.502
Initial r.	9	0.496	0.033	0.028	0.436	3.295	0.084	0.078	0.856	0.565	0.05	0.052	0.384
	10	0.222	0.024	0.028	0.205	0.833	0.054	0.064	0.695	1.199	0.028	0.03	0.193
_	11	0.073	0.007	0.008	0.042	1.096	0.013	0.024	0.503	0.165	0.019	0.022	0.072
Test a	12	0.259	0.009	0.015	0.198	0.589	0.093	0.077	0.17	0.064	0.046	0.046	0.073
Int. r.	13	0.092	0.085	0.078	0.11	0.295	0.211	0.191	0.174	0.167	0.084	0.077	0.224
	14	0.145	0.077	0.073	0.117	0.104	0.141	0.14	0.094	0.198	0.076	0.077	0.232
	15	0.3	0.029	0.032	0.126	0.999	0.069	0.083	0.256	0.133	0.056	0.059	0.306
Advanced r.	16	0.346	0.071	0.049	0.069	0.935	0.179	0.098	0.356	0.345	0.043	0.033	0.114
	17	0.493	0.039	0.09	0.276	1.132	0.164	0.222	0.645	0.364	0.028	0.06	0.274

P = plot; Initial r. = initial regeneration; Int. r. = intermediate regeneration; Advanced r. advanced regeneration.

The Radarsat-2 image in its different polarizations and locations of the 17 plots studied in the field is shown in Figure 3. The observed spatial patterns in the Radarsat-2 HV image generally follow the pattern described by Souza-Filho and Paradella [21] with strong and low backscatter in regeneration and cleared areas, respectively. The co-polarized (*i.e.*, HH and VV) images showed less distinction between those vegetation types. Kovacs *et al.* [17] reported that co-polarized scattering could not be used to distinguish healthy from dead mangroves. It was also observed that the high backscatter from healthy stands is related to very high crown volume scattering from the canopy (branches and leaves), while backscatter from dead and regenerating mangroves is dominated by a double-bounce scattering mechanism from standing water below the canopy acting with the trees as corner reflectors.

The predominance of higher signal returns in the central portion of all of the images (Figure 3) suggests a greater presence of vegetation, which contributes to the occurrence of double-bounce scattering, as a result of trunk-ground interactions during high tides that reach 6 m in range. In addition, C-band images present a higher sensitivity to canopy components which substantially increases scattering at the canopy surface in addition to volume scattering.



Figure 3. Radarsat-2 image in the four polarizations with the locations of the plots studied in the field.

3.2. Analysis of Canopy Structure in Regenerating Mangroves

The total area of the studied plots was 1700 m² in which 2510 live individuals of *A. germinans*, 261 individuals of *L. racemosa*, and 30 individuals of *R. mangle* were measured in addition to 289 dead individuals for a total of 3090 individuals.

An ANOVA analysis was performed to evaluate the similarity between the four groups (recent stage, initial regeneration, intermediate regeneration and advanced regeneration) based on the average of the different structural attributes. Among these attributes, only density did not show significant differences (Table 3). The *post-hoc* Tukey's test showed that BA and AGB had the most significant difference among the four groups. In Table 4, which contains the structural attributes and respective averages separated by stage, the differences between the groups are clear, especially for the attributes BA and AGB. In relation to specific composition, *A. germinans* and *L. racemosa* occur in all plots; however, *A. germinans* is dominant. *R. mangle* occurred only in the advanced regeneration stage (Group IV).

Structural Parameters		Intercept	Group
Lorev's Height	F	43.6397	11.0652
Lorey 5 Height	<i>p</i> -value	0.0000	0.0007
Moon Hoight	F	57.8874	14.1051
Weatt Tiergitt	<i>p</i> -value	0.0000	0.0002
Max Height	F	58.7973	10.2756
Max. Height	<i>p</i> -value	0.0000	0.0010
	F	128.5013	12.0889
DBH	<i>p</i> -value	0.0000	0.0005
DA	F	612.1687	80.2152
ВА	<i>p</i> -value	0.0000	0.0000
D'	F	188.3453	36.4730
BIOMASS	<i>p</i> -value	0.0000	0.0000
Donsity	F	26.1716	1.0950
Density	<i>p</i> -value	0.0002	* 0.3861

Table 3. Analysis of variance (ANOVA) of the structural parameters considering the groups formed in the cluster analysis of the plots (p > 0.05).

* no significant p = 0.05.

Table 4.	Structural	attributes	of the	plots	showing	the fo	rmed grou	ps.
					()		()	

_	6	Dominant	Der	sity	Total	Basal	Mean	Lorey's	Height	Max.	Total
Р	Group	Species	$(ind \cdot ha^{-1})$	(ind ha ⁻¹)	Density	Area	DBH		Mean		Biomass
		-	DBH < 4 cm	DBH > 4 cm	(ind ha ⁻¹)	$(m^2 \cdot ha^{-1})$	(cm)	(m)	(m)	(m)	$(kg \cdot m^{-2})$
1	Ι	Avicennia	1600	-	1600	0.15	1.06	0.32	0.30	0.46	0.03
2			8100	-	8100	1.13	1.21	0.50	0.37	1.13	0.28
3			14,800	-	14,800	1.45	1.04	0.49	0.37	1.04	0.32
4			13,200	700	13,900	3.72	1.55	0.75	0.46	1.43	1.09
5			19,100	400	19,500	5.04	1.63	0.61	0.47	1.4	1.43
6			39,900	400	40,300	5.75	1.18	0.64	0.38	1.8	1.60
	Group a	verage	16,117	500	16,367	2.87	1.28	0.55	0.39	1.21	0.79
7	Π	Avicennia	30,900	600	31,500	7.26	1.50	0.85	0.55	2.37	2.10
8			14,300	1500	15,800	8.31	2.28	1.39	0.93	3.1	2.62
9			10,100	1600	11,700	8.11	2.58	2.81	1.91	4.95	2.69
10			12,800	2000	14,800	10.43	2.62	2.63	2.03	4.78	3.36
11			7300	2700	10,000	10.98	3.20	2.43	1.57	4.55	3.82
	Group a	verage	15,080	1680	16,760	9.02	2.44	2.02	1.39	3.95	2.92
12	III	Avicennia	52,500	1200	53,700	14.56	1.63	1.95	1.18	4.13	4.35
13			4100	3100	7200	14.11	4.34	2.43	1.96	4	5.43
	Group a	verage	28,300	2150	30,450	14.34	2.98	2.19	1.57	4.07	4.89
14	IV	Avicennia	13,100	4700	17,800	20.42	3.38	6.95	5.08	10.85	6.50
15			5200	3800	9000	18.34	4.35	3.18	2.09	4.74	7.36
16			2400	4300	6700	20.89	5.63	4.92	3.88	7	9.53
17			1600	2100	3700	20.78	6.90	10.55	6.39	15.15	11.24
	Group a	verage	5575	3725	9300	20.11	5.07	6.40	4.36	9.44	8.66

3.3. Estimating Structural Attributes of Regenerating Mangrove Vegetation from SAR Data

The best correlation with structural attributes was found for cross-polarization backscatter (Table 5). The VV polarization obtained low and inverse correlations, which can be related to the lower stature of the vegetation. This result may occur because we are working with a regenerating forest where horizontal scattering predominates and causes the opposite of what was described by van der Sanden [24].

	Lorey's Height	Mean Height	Max Height	DBH	Basal Area	Total Biomass
	(m)	(m)	(m)	(cm)	$(m^2 \cdot ha^{-1})$	(kg \cdot ha $^{-1}$)
$\beta^{\circ}_{\rm HH}$	0.20	0.17	0.22	0.22	0.15	0.18
$\beta^{\circ}_{\rm HV}$	0.56	0.60	0.54	0.61	0.62	0.64
β°_{VH}	0.71	0.70	0.69	0.67	0.62	0.68
β°_{VV}	-0.74	-0.12	-0.08	-0.14	-0.17	-0.13
$\sigma^{\circ}_{ m HH}$	0.56	0.53	0.56	0.52	0.47	0.55
$\sigma^{\circ}_{\rm HV}$	0.51	0.60	0.49	0.58	0.62	0.58
σ°_{VH}	0.77	0.79	0.75	0.73	0.68	0.72
σ°_{VV}	-0.03	-0.07	-0.02	-0.16	-0.14	-0.12
$\gamma_{\rm HH}$	0.20	0.24	0.23	0.19	0.12	0.12
$\gamma_{\rm HV}$	0.35	0.42	0.37	0.40	0.59	0.45
γ_{VH}	0.58	0.61	0.59	0.55	0.66	0.59
γνν	0.03	0.01	0.04	0.00	-0.05	-0.04

Table 5. Correlation coefficient between structural attributes and backscattering of the Radarsat-2 image FQ5. The highest correlation coefficient values are highlighted (p < 0.05).

Different functions were fitted to the set of variables with significant correlation coefficients. Table 6 shows that the best fit of the regression function is linear and best with the σ°_{VH} backscattering.

Table 6. Models that showed higher r^2 values in the three radar attributes with VH polarization as an explanatory variable (p > 0.05).

Lorey's Height							Mean Height				Maximum Height				
		r ²	β1 (p)	F	р		r ²	β1 (p)	F	р		<i>r</i> ²	β1 (p)	F	р
σ°_{VH}	LIN	0.59	0.000	21.909	0.000	LIN	0.63	0.000	25.237	0.000	LIN	0.57	0.000	19.587	0.000
β°_{VH}	LIN	0.50	0.002	14.795	0.002	LIN	0.49	0.002	14.612	0.002	LIN	0.48	0.002	13.681	0.002
γνн	EXP	0.41	0.006	10.265	0.006	EXP	0.43	0.004	11.42	0.004	EXP	0.37	0.009	8.958	0.009
			D	BH			Basal Area					Biomass			
		r^2	β1 (p)	F	р		r^2	β1 (p)	F	р		r^2	β1 (p)	F	р
σ°_{VH}	LIN	0.53	0.001	16.979	0.000	LIN	0.46	0.003	12.950	0.003	LIN	0.52	0.001	16.430	0.001
$\beta^{\circ}vh$	LIN	0.44	0.003	11.987	0.000	LIN	0.39	0.008	9.466	0.008	LIN	0.46	0.003	12.572	0.003
γ_{VH}	EXP	0.35	0.013	7.944	0.013	LIN	0.44	0.014	11.671	0.004	LIN	0.35	0.012	8.073	0.012

Multiple linear regression models were subsequently fitted to potentially increase the predictive power of the regressions. The multicolinearity between the independent variables that compose these models was verified by VIF (variance inflation value), which resulted in the values of 1.41, 1.13 and 1.27 for σ°_{HH} , σ°_{VH} , and σ°_{VV} , respectively. These values are below the limit value of 10 indicated by Neter *et al.* [50]. The parameters of these models are provided in Table 7.

The variable σ°_{VV} (β_3 , Table 7) was not statistically significant in the regression model of the attribute maximum height previously described. Therefore, it is possible that the vertical components of the vegetation are not sufficiently developed to interact with microwaves in the VV polarization.

In the residual analysis, the models met the assumptions proposed by Neter *et al.* [50]. With the PRESS values, the models produced adequate values, especially those for horizontal structures, which showed a better predictive ability in the fit regression function with an emphasis on the DBH model (Table 8).

Although the maximum height model was simpler (Figure 4), the average height model had a higher predictive ability based on the RMSE value. For the estimation of the horizontal structure, the model for DBH had the best predictive ability, although other models were satisfactory (Figure 4). When comparing the verified modeling methods, r^2 values increase with the introduction of σ°_{HH} and σ°_{VV} backscattering as independent variables. The explanatory power increased between 11% and 19% for the models of height estimation and between 20% and 27% for the models of estimation

of horizontal structure and AGB. The RMSE values decreased with the inclusion of these variables; the PRESS values also decreased. This indicates that these models should be chosen instead of the simple regression models [51].

	Lorey's	Height	Mean	Height]	Max. Height	:		
	Simple	Multiple	Simple	Multiple	Simple	Mul	tiple		
r ²	0.59	0.79	0.63	0.81	0.57	0.76	0.68		
β ₀	0.020	-0.515	0.049	-0.191	0.817	0.010	-0.593		
$\beta_{1(\sigma^{\circ}HH)}$	-	0.502	-	0.468	-	0.505	0.636		
р ́	-	0.006	-	0.007	-	0.008	0.001		
$\beta_{2(\sigma^{\circ}VH)}$	0.770	0.641	0.792	0.677	0.753	0.621	4.323		
p ´	0.000	0.000	0.000	0.000	0.000	0.001	0.045		
$\beta_{3(\sigma^{\circ}VV)}$	-	-0.329	-	-0.356	-	-0.318	-		
p í	-	0.040	-	0.022	-	0.059	-		
ε	1.7894	1.396	1.1248	0.870	2.5923	2.082	2.312		
F	21.909	15.869	25.237	18.087	19.587	13.541	14.743		
р	0.000	0.000	0.000	0.000	0.000	0.000	0.000		
	DI	BH	Total B	iomass		Basal Area			
	Simple	Multiple	Simple	Multiple	* Simple	** Simple	Multiple		
r^2	0.53	0.79	0.52	0.79	0.46	0.50	0.67		
β ₀	0.473	1.033	87.056	39.009	0.021	0.007	0.038		
$\beta_{1(\sigma^{\circ}HH)}$	-	0.531		1278	-		0.472		
р ́	-	0.004		0.003	-		0.027		
$\beta_{2(\sigma^{\circ}VH)}$	0.729	0.604	0.723	7279	0.681	0.614	0.571		
p ´	0.001	0.001	0.001	0.001	0.003	0.001	0.005		
$\beta_{3(\sigma^{\circ}VV)}$	-	-0.465		-868	-		-0.422		
p í	-	0.007		0.009	-		0.035		
ε	1.2193	0.882	234.83	168.110	0.0534	0.0514	0.045		
F	16.979	16.033	16.430	16.109	12.950	15.150	8.788		
p	0.001	0.000	0.001	0.000	0.003	0.001	0.002		

Table 7. Parameters of the simple and multiple regression models and $\sigma^{\circ}_{HH_{r}} \sigma^{\circ}_{VH_{r}}$ and σ°_{VV} backscattering that compose the independent variables.

Legend: r^2 = determination coefficient; β_0 = line intercept; $\beta_{1,2,3}$ = line inclination; ε = random error, F = Fischer test for total variance model, * σ°_{VH} (5 × 5) = independent variable, and ** γ°_{VH} (3 × 3) = independent variable.

Table 8. PRESS values and SQR difference percentage.

		Multiple		Simple				
Attribute	PRESS	SQR	%	PRESS	SQR	%		
Lorey's Height	53.59	25.35	52.69	79.47	48.03	39.56		
Mean Height	18.50	9.84	46.82	28.61	18.98	33.68		
Max. Height	140.51	74.83	46.74	162.38	100.80	37.92		
DBH	15.69	10.12	35.53	29.69	22.30	24.89		
Total Biomass	639,023	367,408	42.50	1,072,887	827,197	22.90		
* Basal Area	0.04	0.03	39.80	0.052	0.043	17.47		
** Basal Area	-	-	-	0.049	0.040	19.26		

* σ°_{VH} (5 × 5): independent variable; ** γ_{VH} (3 × 3): independent variable.



Figure 4. Plots of the observed values against the predicted values, with respective r^2 and RMSE values.



Figure 5. Estimation map: (A) average DBH (cm) and (B) average height (m) and (C) total biomass $(kg \cdot m^{-2})$ based on the backscattering values through their multiple regression functions.

The fitted regression models were developed and validated, and then applied to the backscattering values from the Radarsat-2 FQ5 image to generate maps of DBH, average height and AGB (Figure 5). The values shown in the average DBH map ranged between 1.2 and 3.3 cm, which is consistent with the data measured in the field, in which only four sample units had values above 3.3 cm. The map showed a few regions with DBH lower than 1.6 cm, and most of the individuals with greater DBH were in the central portion of the map and ranged from 2 to 3.3 cm. The applied parameter was the average DBH, whose model RMSE was 0.77 cm; because it is a regenerating mangrove region, the amplitude of variation of this measurement is high as a result of the structural heterogeneity. The average height ranged from 0.2 to 1.9 m and is considered consistent with the values measured in the field, especially when considering the RMSE of the model, which was 0.76 m; there were only three plots outside of this height range.

The total AGB map showed a large value variation between 0 and 60 kg·m⁻². These values include all AGB measured in the field. Zero represents areas without vegetation with exposed tidal flats. The more frequent values are between 10 and 40 kg·m⁻². This model seemed to overestimate AGB, which is most likely a result of double-bounce scattering.

4. Conclusions

The regenerating mangrove vegetation showed structural heterogeneity with a wide range of structural parameter variation, and the BA was the best variable to distinguish the regeneration stages. Four stages were differentiated into groups: recent stage (Group I), initial regeneration (Group II), intermediate regeneration (Group III) and advanced regeneration (Group IV). The dominant species in the greatest number was *Avicennia germinans*. The species *Laguncularia racemosa* had the lowest occurrence in the four groups and the species *Rhizophora mangle* was only found in the advanced regeneration group. The equation used to calculate AGB reflected the high range of variation between the four groups and can be considered adequate. Particularly, linear sigma backscattering σ° showed the strongest and most significant correlation with the structural data from the regenerating mangrove vegetation, especially in the VH cross-polarization.

The multiple regression model with the σ°_{HH} , σ°_{VH} and σ°_{VV} polarization showed high predictive capacity for the variables' average height ($r^2 = 0.81$), DBH ($r^2 = 0.79$) and AGB ($r^2 = 0.79$), which permitted the generation of maps of these vegetation attributes. Therefore, DBH and average height maps exhibit values commensurate with those observed in the fieldwork. The central region of the study site showed the highest values of DBH and average height, and consequently, this region showed the highest values of total AGB. The AGB measured in the field presented a high correlation with Radarsat-2 backscattering. Finally, this study provided important new insights into the interpretation of multi-polarized Radarsat-2 images, which showed to be adequate for the estimation of vegetation attributes in areas of mangrove regeneration. Additional research will explore the influence of full polarimetric C-band RADARSAT-2 data (decomposition and polarimetric response), involving all successional stages of mangrove vegetation.

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fieldwork to collecte and analyze geo-referenced structural mangrove data. Marc Simard helped with structural parameters and biomass estimation. All authors read and approved the final version of the manuscript.

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