

# ABOVE GROUND BIOMASS FUNCTIONS WITH VEGETATION INDICES FOR MULTIPLE USE SYSTEMS OF TWO EVERGREEN OAKS

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## INTRODUCTION

The aim of this study is the development of allometric functions of above ground biomass that can be used in both monospecies and multispecies stands of the two evergreen oaks, whose independent variable are vegetation indices derived from high spatial resolution satellite image (QuickBird). As some differences can be found between the stand parameters of monospecies stands of cork oak or holm oak and multispecies stands of these two evergreen oaks, it was also considered the development of allometric functions where stand composition is included as dummy variables.

## MATERIAL AND METHODS

The QuickBird image (August, 2006) with a spatial resolution of 0.70 m, had the following main processing steps:

- geometric correction using ground control points, atmospheric correction, conversion of the digital numbers to ToA reflectance;
- the vegetation mask, resulted from the application of object-oriented classification image methods;
- vegetation indices were calculated based on the individual bands of the satellite image. The vegetation index value per grid was calculated as the arithmetic mean and the median of all pixels within each grid.

Other methodological steps were:

- the division of the image in a square grid of 45.5 x 45.5 m (2070.25 m<sup>2</sup>) with the same dimension of the field plots;
- random stratified sampling by proportional allocation design was used in forest inventory, where each plot corresponds to a grid. Tree above ground biomass was calculated using the plot data and the functions of Paulo and Tomé (2006);
- the use of linear and multiple regression to fit the functions of above ground biomass (AGB) with vegetation indices as independent variable;
- plot species composition was defined as a dummy variable where: monospecies of cork oak (*dQS*), monospecies holm oak (*dQR*), and multispecies of these two evergreen oaks (*dQRQS*).

The linear regression was fitted with the ordinary least squares, and multiple linear regressions with stepwise method and the selection criteria AIC using Variance Inflation Factor (VIF) to evaluate multicollinearity among explanatory variables.

The statistical properties of the models and validation were done with: the sum of squares of the residuals (SQR), determination coefficient (R<sup>2</sup>) and adjusted determination coefficient (R<sup>2</sup><sub>aj</sub>).

The validation was done using PRESS and APRESS. The error term heteroscedasticity was assessed by plotting the studentised residuals vs estimated values. The normality of the studentised residuals was evaluated with the normal Quantile-Quantile plots (QQ-plots) and the Shapiro Wilk normality test, for a probability level of 0.001.

## RESULTS

### Plots descriptive statistics

	Min	Max	Average	SD	CV (%)
All plots					
AGB (t/ha)	10.4	62.7	28.5	10.4	36.4
GC (%)	13.7	70.5	34.3	14.9	43.6
Monospecific cork oak plots					
AGB (t/ha)	23.8	62.7	39.9	12.9	32.4
GC (%)	23.6	70.5	47.1	14.0	29.7
Monospecific holm oak plots					
AGB (t/ha)	13.7	67.6	36.5	16.7	45.7
GC (%)	18.7	45.9	27.9	8.4	29.9
Multispecies holm oak and cork oak plots					
AGB (t/ha)	10.4	35.2	23.9	6.2	25.9
GC (%)	14.3	45.3	26.9	9.0	33.3

### Models with best results for each plot type and statistic measure used

Model	Ind. Var	Per plot				Per vegetation mask					
		SQR	R2aj	PRESS	APRESS	Model	Ind. Var	SQR	R2aj	PRESS	APRESS
Holm oak monospecies											
M1	NDVI	250	0.725	0.01876	0.46757	M9	NDVI	194	0.769	0.01853	0.43273
M7	SR	276	0.697	0.01743	0.44702	M15	SR	181	0.785	0.01854	0.43718
Cork oak monospecies											
M17	NDVI	515	0.559	0.00586	0.18576	M26	EVI	437	0.563	0.00589	0.17820
M23	SR	480	0.589	0.00555	0.17794	M32	SAVI	451	0.550	0.00589	0.17922
Holm oak and cork oak multispecies											
M34	EVI	296	0.570	0.06756	0.86807	M41	NDVI	354	0.457	0.07238	0.85708
M38	EVI	296	0.570	0.06756	0.86807	M45	NDVI	344	0.471	0.07238	0.85708
All plots											
M51	SR	1213	0.733	0.06589	1.33629	M59	SR	1354	0.694	0.07545	1.33594
M55	SR	1389	0.694	0.06589	1.33629	M63	SR	1372	0.690	0.07545	1.33594
All plots with dummy variables											
M65	NDVI, dQR, dQS	1150	0.740	0.06321	1.34521	M75	SR, dQR, dQS	2694	0.377	0.08599	1.37854
M71	SR, dQR, dQS	1156	0.739	0.07327	1.36187	M79	SR, dQR, dQS	2734	0.368	0.08693	1.37385

## CONCLUSIONS

The best models were obtained for holm oak monospecies plots, these models presented lower SQR, PRESS and APRESS values and highest adjusted determination coefficient value, the models for all plots without accounting for vegetation mask also returned good results.

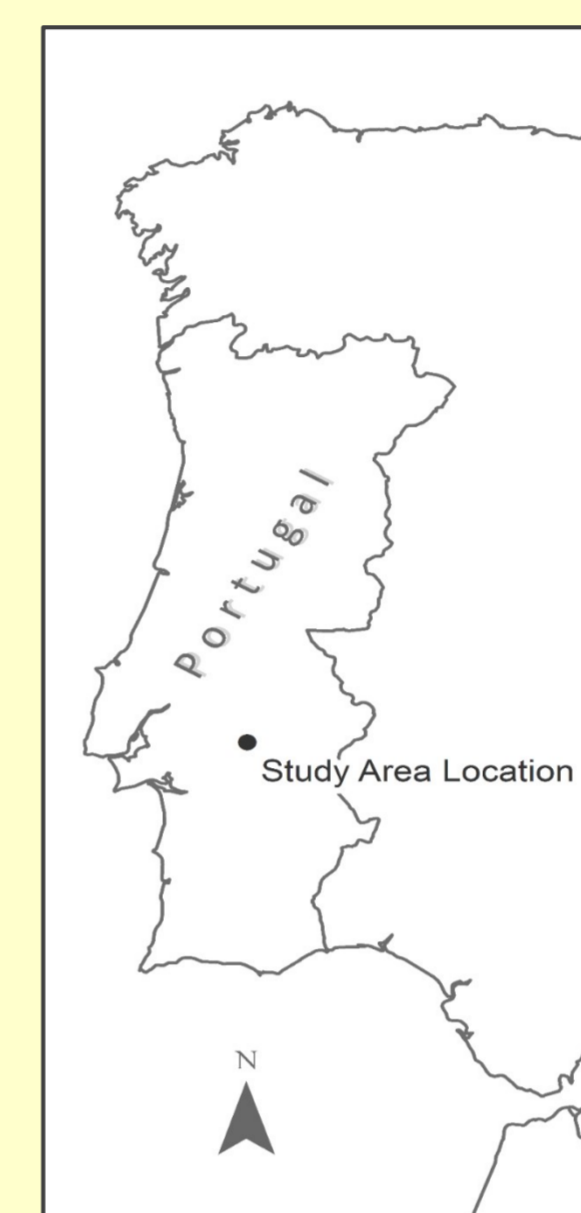
The vegetation mask results show that all the models are very similar within plot composition, revealing that different indices and statistical measures produced very redundant models.

SR and NDVI were the indices with more frequency of overall better results, it was not noticed a relevant pattern in the difference between mean or median usage models, with overall similar results being obtained.

The inclusion of plot composition slightly increases the models performance per plot and a strongly decreases the performances of the models per vegetation mask.

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$$AGB = \beta_0 + \beta_1 \times VI + \beta_2 \times dVI$$

Vegetation Index	Equation	Reference
Normalized Difference Vegetation Index	$NDVI = \frac{NIR - RED}{NIR + RED}$	(Rouse et al., 1973)
Enhanced Vegetation Index	$EVI = \frac{2.5 \times (NIR - RED)}{(NIR + 6 \times RED - 7.5 \times BLUE + 1)}$	(Huete et al., 1996; 1997)
Simple Ratio	$SR = \frac{NIR}{RED}$	(Jordan, 1969)
Soil Adjusted Vegetation Index	$SAVI = \frac{(NIR - RED)}{(NIR + RED + L)} \times (L + 1)$	(Huete, 1988)

