

Bases para la planificación sostenible de áreas marinas en la Macaronesia

Identification of areas for Blue Growth











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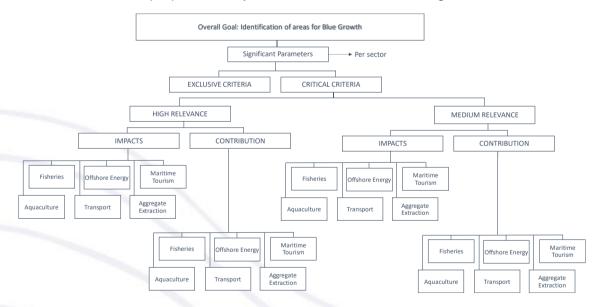
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Executive Summary

This report will describe a proposal of a decision-making support tool that aims to achieve a hierarchy definition for contribution and impact of parameters for different maritime sectors in Macaronesia. The methodology is proposed under the PLASMAR activity 2.1.2 "Pilot Zoning – identification of areas for Blue Growth development within ecosystem approach" and will contribute to the spatial identification of areas for blue growth through the parameters' grading system.

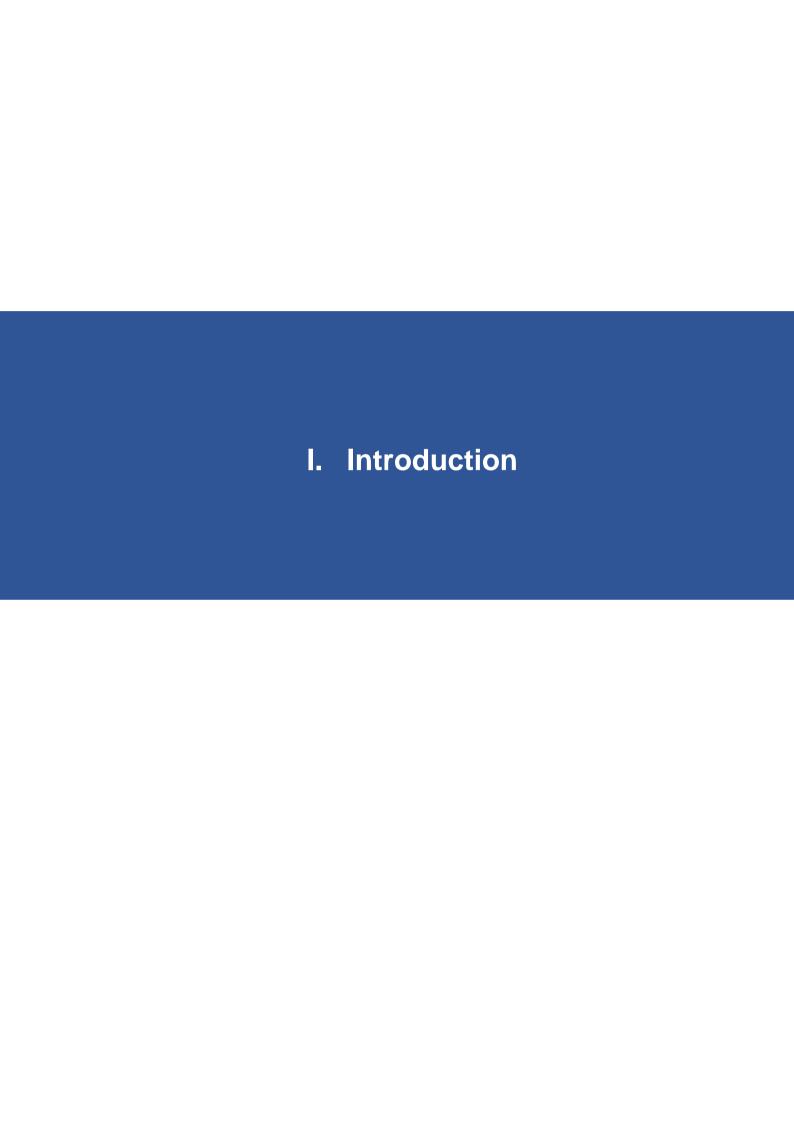
The proposal consists on the adaptation of a Multicriteria Decision Analysis methodology, the Analytical Hierarchy Process (AHP). The AHP is one of the most widely used multicriteria decision-making technique and its use is based on relatively easy procedures as well as on the possibility to evaluate the inconsistency (Aguarón, Escobar and Moreno-Jiménez, 2016). Moreover, the AHP analysis is mainly based on: the hierarchy of interrelated decision elements describing the problem, composing a decision-tree; the pairwise comparison of the decision elements; and the consistency ratio.

In this methodological proposal, the criteria analysed were defined as the PLASMAR Parameters, and the pairwise comparison aims to be based on experts' consultation. The structure of the proposed analysis can be observed in the Figure below.



Final proposition of hierarchical structure for the PLASMAR analysis.

As observed in the Figure above, the analysis will consist on the selection of PLASMAR significant parameters, the significance will be based on the attributed high and medium relevance defined within the project. Following this first selection, the parameters contribution and sectorial impacts over the parameters will be analysed to each maritime sector trough the multicriteria methodology. The final output of the analysis will be a score to each parameter per sector. For further details, a step-by-step approach to guide the methodological proposal application was developed in the report.



1 Introduction

Aiming the development of methodologies to support Maritime Spatial Planning (MSP) and Blue Growth, the project PLASMAR "Setting the basis for sustainable maritime spatial planning in Macaronesia" has the goal to potentialize the development of marine activities in balance with the biogeographical specific characteristics of Macaronesia region (including in Portugal the Azores and Madeira archipelagos, and in Spain the Canary Islands). It also aims to support the achievement of the Good Environmental Status (GES) launched at the Marine Strategy Framework Directive (MSFD) (Directive 2008/56/EC).

The PLASMAR activity 2.1.2 "Pilot Zoning – identification of areas for Blue Growth development within ecosystem approach" intends to identify potential areas for the "blue growth" in Macaronesia. This task consists on developing a general methodology of zoning activities/sectors over the maritime space and implementing a pilot zoning for Macaronesia, which will be developed on INDIMAR platform.

The elaboration of a zoning proposal demands basis information. Hence, this data will be gathered, organized and produced in the following activities:

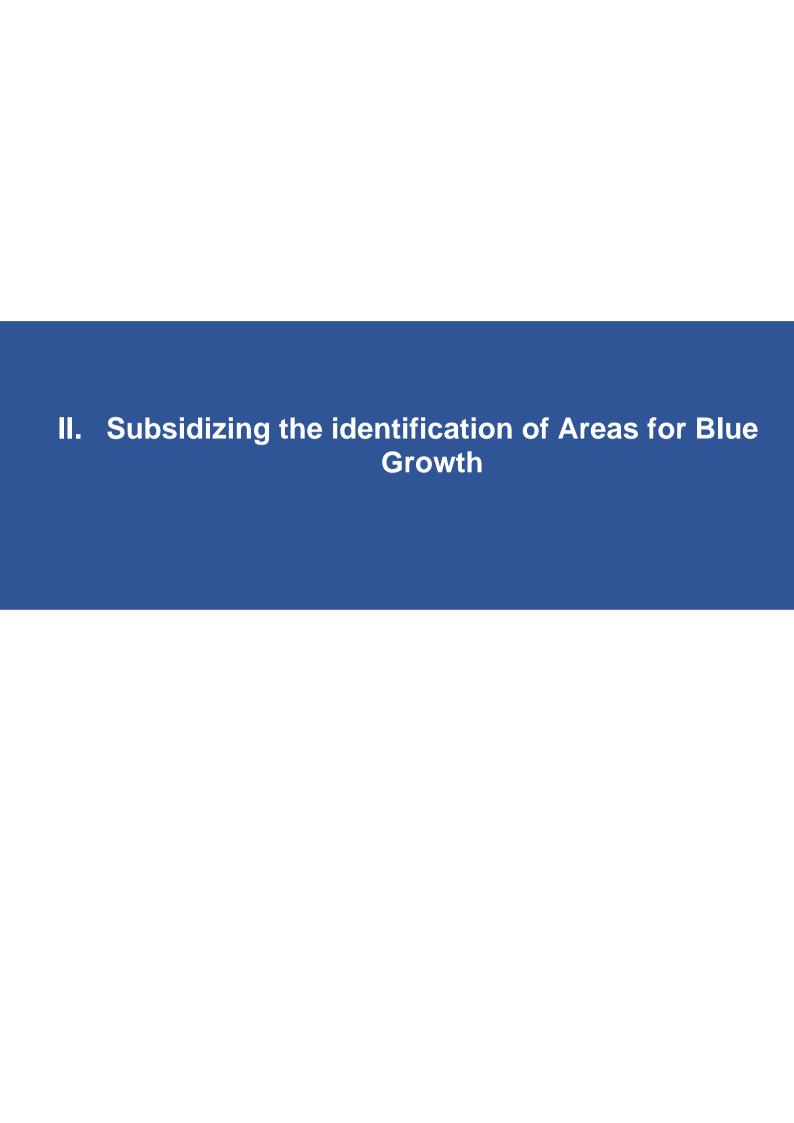
- i. Activity 2.1.1 "Finding the balance of Blue Growth Sustainable Development within Ecosystem Approach";
- ii. Activity 2.2.1 "Analyses of data & information availability, current and potential data holders/providers, in the scope of Maritime Spatial Planning";
- iii. Activity 2.3.1. "Marine monitoring methods needed to apply MSP ecosystem approach".

From the results obtained in the project, a zoning methodology will be stablished considering the following marine sectors:

- i. Aquaculture;
- ii. Fisheries;
- iii. Marine transportation;
- iv. Offshore renewable energy;
- v. Aggregate extraction;
- vi. Marine tourism.

The pilot zoning will identify the best suitable areas for specific maritime activities, in line with the maintenance of the GES, according to the information available in the Marine Distributed Data Infrastructure. This is a result of the Activity 2.2.1 of PLASMAR "Analyses of data & information availability, current and potential data holders/providers, in the scope of Maritime Spatial Planning".

In this sense, the different methodologies currently applied for Multi-Criteria Decision Making (MCDM) will be further developed in order to subsidize the methodology currently being developed in this project.



2 Multicriteria Decision Analysis

The concept of Multicriteria Decision Making (MCDM), or Multicriteria Decision Analysis (MCDA), concerns structuring and solving decision and planning problems involving multiple criteria (Majumder, 2015). Steps to develop a MCDA, in a participatory way, are represented in Figure 1 (Durham *et al.*, 2014).

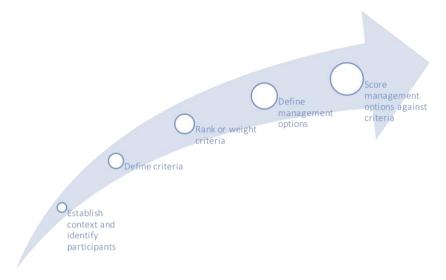


Figure 1:Steps of MCDA. Source: Adapted from (Durham et al., 2014).

The concept of MCDM concerns problems that demand one decision between several possible alternatives influenced by different criteria. The main methodologies are listed by (Velasquez and Hester, 2013):

- 1. ANP Analytic Network Process
- 2. AHP Analytic Hierarchy Process
- 3. MAUT Multi-Attribute Utility Theory
- 4. TOPSIS Technique for Order of Preference by Similarity to Ideal Solution
- 5. PROMETHEE Preference Ranking Organization Method for Enrichment of Evaluations
- 6. ELECTRE Elimination and Choice Expressing Reality
- 7. SAW Simple Additive Weighting
- 8. MACBETH Measuring Attractiveness by a Categorical Based Evaluation Technique

The application of these methods will vary depending on the application area, due their specific characteristics. Huang, Keisler and Linkov, 2011 explore the use of each method in the field of environmental sciences based on the analysis of 300 study cases between 2001 and 2009. Similar study was taken by (Marttunen, Lienert and Belton, 2017), the authors analysed 68 papers published between 2000 and 2015. In Figure 2 the representation of the results of both articles is represented through the most common uses in different areas and the percentage of application of each method.

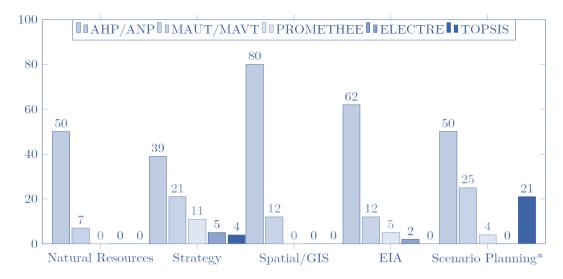


Figure 2: Use of MCDM methods in different fields of science (based on Huang et al. 2011).

From Figure 2, the AHP/ANP method has being largely used in literature and across several scientific fields. Furthermore, it is important to observe that other methods express moderate relevance in application.

In a methods comparison (Table 1), it is possible to observe that AHP presents good precision with low amount of data, time and technical knowledge. Moreover, even though the results are not as precise for the AHP methodology as for the other methods, the consistency checking guarantees the reliability of the analysis and builds a good problem modelling possibility with limited data. In MSP, as well as small islands context as the ones addressed by the PLASMAR project, data acquisition can be costly or even impracticable. As a result, AHP can be a powerful resource for these realities, based on the knowledge of experts, technicians and professionals in the area.

Table 1: Models comparison.

Criteria	AHP	MAUT	TOPSIS	
Precision	Good	High	Good	
Versatility	High	Very high	High	
Amount of data	Low	High	High	
Time	Low	High	Low	
Technical knowledge	Low	High	Medium	
Specific software	No	Yes	Programmable	
Method	Pairwise comparison	Utility function	Ideal and anti-ideal options	
Output	Punctuation ranking	Punctuation ranking	Partial ranking with efficacy score	

2.1 Analytical Hierarchy Process (AHP)

The Analytical Hierarchy Process (AHP) is a general theory of measurement (Saaty, 1987) and a useful decision-making support technique (Soma, 2003). The AHP is one of the most widely used multicriteria decision-making technique for dealing with multi/stakeholders problems (Aguarón, Escobar and Moreno-Jiménez, 2016). Its use is based on relatively easy procedures and on the possibility to evaluate the inconsistency of the preferences (Aguarón, Escobar and Moreno-Jiménez, 2016).

The AHP methodology involves four main steps (Saaty, 1990). The steps based on Soma (2003) description, and the actions taken in this study to cover them will be further explained:

1. Developing a hierarchy of interrelated decision-elements describing the problem (Figure 3). To address this step, the current study dedicated the sections "Establishing the context" (Sections 3.1, 3.2, 3.3).



Figure 3: Example of decision-tree. Source: Soma, 2003.

- 2. Performing a pairwise comparisons of the decision elements through a weighting scale to generate input data, with all the decision elements agreed upon by all the stakeholder groups. This step is comprehended in the section 3.3 "Specifying preferences", specifically in section 3.3.1 "Step 1: Determination of the weight coefficients for the criteria".
- 3. Computing the relative weights of the decision-elements. This step is comprehended in the section 3.3 "Specifying preferences", specifically in section 3.3.2. "Step 2: Calculating the weights per sector".
- 4. Determining the prioritisation of objectives, criteria and management options in order to identify the alternative decision possibilities. This step is partially comprehended in the section 3.3 "Specifying preferences", specifically in section 3.3.4. "Step 4: Calculating the final weight". This study aims to compare the different criteria, however, not the alternatives. Further details are given in the next sections.

3 Identifying areas for Blue Growth: Defining weights to parameters

Aiming the achievement of the Action 2.1.2. of PLASMAR, "Pilot Zoning – Identification of areas for Blue Growth development within ecosystem approach", a detailed methodology of weight calculation for the parameters identified under the project will be further proposed in order to subsidize the goal of the action.

3.1 Establishing the context: Defining the goal and the objectives

Initially, in order to guide the research and the pairwise comparison analysis, the overall goal of the research must be defined (Saaty, 1990). Considering the PLASMAR action 2.2.1, on which this work is based, the overall goal considered on this analysis was the "Identification of areas for Blue Growth", as observed in Figure 4. Moreover, this goal was subdivided in order to comprehend the relation of the different parameters already developed by the project. The subdivision consisted on "Contribution", referring to the parameters contribution to the sector, and the "Impacts", referring to the parameters impacted by the sector.

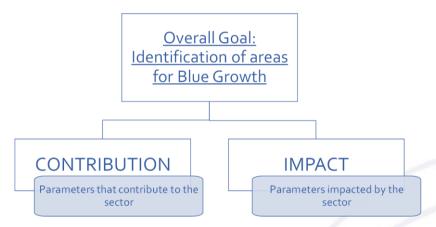


Figure 4: Goal definition in the analysis.

Secondly, in order to comprehend the different maritime sectors addressed during the project as the blue growth sectors, the overall goal, and consequently, each goal's subdivision, was further subdivided into objectives. This subdivision aimed the identification of specific criteria and weights for each sector (*Figure 5*).

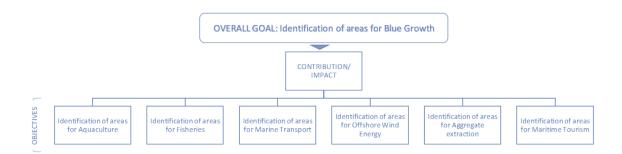


Figure 5: Objectives definition

3.1.1 Establishing the context: Defining the criteria

The criteria adopted for the multicriteria analysis in this study were the PLASMAR data framework, in order to associate the aforementioned analysis with the spatial data developed in the scope of the project. The structured list of clusters of criteria (group of parameters or cluster), criteria (parameters) and sub-criteria (sub-parameters) is presented in Table 2. Details on the selection of the parameters will be available soon at plasmar.eu. Moreover, the criteria still under partners' evaluation and, as a result, is not yet fully developed. This list might need to be updated on the analysis.

Table 2: PLASMAR data framework adopted for the identification of areas for Blue Growth.

Group of parameters	Parameter	Sub-parameter Marine Habitats		
		Coastal habitats		
	December 1. Die diversity	Sensitive species distribution/migrations		
	Descriptor 1: Biodiversity	Mammals		
	 Habitats and Species 	Birds		
		Turtles		
MSFD Good		Cephalopod		
Environmental	Descriptor 2	Non-indigenous species		
Status criteria	Descriptor 3	The population of commercial fish species		
elements	Descriptor 4	Elements of food webs		
		Chlorophyll a		
	Descriptor 5: Human-	Dissolved oxygen		
	induced eutrophication	Nutrients		
		Water transparency		
	December 6	The sea floor integrity (physical loss &		
	Descriptor 6	disturbance)		

	Descriptor 7	Permanent alteration of hydrographical conditions			
	Descriptor 8	Concentrations of contaminants – heavy metals and other contaminates			
	Descriptor 9	Contaminants in seafood - assessed data, not time series			
	Descriptor 10	Marine litter			
	Descriptor 11	Energy, including underwater noise data			
	Nationally designated area				
MPA	Natura 2000	· · · ·			
	 Local data				
	CORINE (land cover)				
	Land use				
Land use/cover	No take zone				
•	Distance to the coast				
	Point and lineal coastal pressures				
	Temperature .				
	Sea surface/sea bottom temperature				
	Sea salinity				
	Mixed layer thickness				
Oceanography	Currents				
	Waves				
	 Depth/bathymetry				
	Wind				
		Aquaculture facilities			
	Aquaculture pressures	Seaweed cultivation			
		Maritime traffic lanes/intensity maps			
	Maritime Transportation	Offshore supply/bunkering			
	pressures	Harbours/ports			
		Fish landings			
	Fisheries pressures	Fishery areas/efforts			
Maritime		Aggregate extraction			
activities/pressures	Mineral Resources	Dredging / Sand extraction			
	pressures	Fossil fuel extraction			
		Deep sea mining			
	1/	Artificial reefs			
		Coastal Tourism			
	Tourism pressures	Cruise Tourism			

	Diving
	Windsurfing
	Wreck
	Dumping
	Desalination
	Cables
Others	Military area
Others	Ship building
	Blue biotech
	Coastal point pressures (submarine outfalls,
	gullyies)
	Research area

In order to evaluate the relevant parameters for each maritime sector, the critical parameters might be identified by expert's consultation, which is currently being developed among the project partners. It is worth referring that the current parameter survey taking place among the partners aims to determine the significance/relevance of the parameters, as well as identify missing parameters. In this survey, the importance of each parameter will be classified as "low", "medium" and "high" for each activity. Thus, the critical parameters will be considered as the ones defined as "high" and "medium".

Later, based on this first selection, a second selection will classify the critical parameters in "exclusive criteria" and "critical criteria" (*Figure* 6). The "exclusive criteria", based on Botelho *et al.* (2015), are defined as the criteria that presents no trade-off for its development due to legal constrains, for instance, the jurisdictional area of the port, buffer areas for diving and recreational activities, marine protected areas (MPAs), etc. On the other hand, the "critical criteria" are the remaining critical parameters classified for each activity, these are the criteria that will be weighted on the multicriteria analysis taken in this study. This selection aimed to build a hierarchy of the interrelated decision-elements through the experts, resulting in criteria with comparable unities. Additionally, "medium" and "high" parameters were separately considered in order to preserve the homogeneity in the pairwise comparison, however, if necessary (e.g., the case of few parameters classifies for "medium" or "high"), they can be jointly considered. Moreover, if only one parameter remains in the cluster after the classification, it is advisable to consider all the parameters in the cluster for the pairwise comparison.

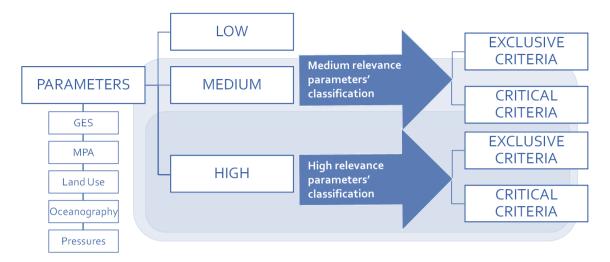


Figure 6: Parameters selection for the multicriteria analysis.

The final hierarchical structure proposed for this analysis can be observed in Figure 7.

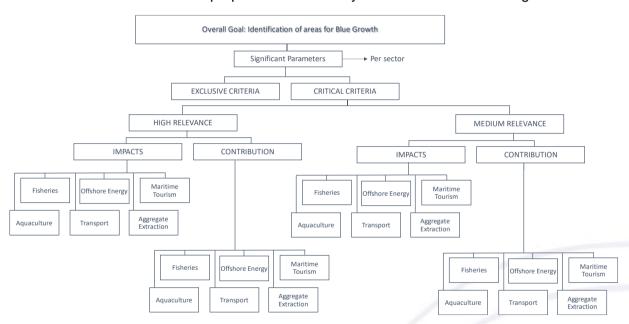


Figure 7: Final proposition of hierarchical structure for the PLASMAR analysis.

3.1.2 PLASMAR Parameters Definition

A brief description of some parameters is given below.

• MSFD Good Environmental Status

Biodiversity	Biological diversity is maintained. The quality and occurrence of habitats and the distribution and abundance of species are in line with prevailing physiographic, geographic and climatic conditions
Non-indigenous species	Non-indigenous species introduced by human activities are at levels that do not adversely alter the ecosystems
Commercial fish species	Populations of all commercially exploited fish and shellfish are within safe biological limits, exhibiting a population age and size distribution that is indicative of a healthy stock
Elements of food webs	All elements of the marine food webs, to the extent that they are known, occur at normal abundance and diversity and levels capable of ensuring the long-term abundance of the species and the retention of their full reproductive capacity
Eutrophication	Human-induced eutrophication is minimised, especially adverse effects thereof, such as losses in biodiversity, ecosystem degradation, harmful algae blooms and oxygen deficiency in bottom waters
The sea floor integrity	Sea-floor integrity is at a level that ensures that the structure and functions of the ecosystems are safeguarded and benthic ecosystems, in particular, are not adversely affected
Permanent alteration of hydrographical conditions	Permanent alteration of hydrographical conditions does not adversely affect marine ecosystems
Concentrations of contaminants	Concentrations of contaminants are at levels not giving rise to pollution effects
Contaminants in Seafood	Contaminants in fish and other seafood for human consumption do not exceed levels established by Community legislation or other relevant standards
Marine litter	Properties and quantities of marine litter do not cause harm to the coastal and marine environment
Energy and noise	Introduction of energy, including underwater noise, is at levels that do not adversely affect the marine environment

• Maritime Protected Areas (Several sources)

Natura 2000	Natura 2000 is a network of core breeding and resting sites for rare and threatened species, and some rare natural habitat types which are protected in their own right. The aim of the network is to ensure the long-term survival of Europe's most valuable and threatened species and habitats, listed under both the Birds Directive and the Habitats Directive
No take zones	Protected areas closed to fishing
CDDA by EEA	The Nationally designated areas inventory (CDDA), formerly known as the Common Database on Designated Areas, is now an agreed annual Eionet core data flow maintained by the European Environment Agency (EEA) with support from the European Topic Centre on Biological Diversity (ETC/BD). The CDDA data is delivered by each country as spatial information (Type 1) and thematic information (Type 2) and the dataset is used by the EEA and e.g. the

- Land Use/Land Cover
- Oceanography (Source: http://marine.copernicus.eu/training/education/oceanparameters/)

Sea Surface Temperature (SST)	Sea surface temperature is the temperature of the ocean near the surface
Sea Bottom Temperature (SBT)	Sea bottom temperature is the temperature of the ocean near the bottom
Currents	By transporting heat and energy, ocean currents play a major role in shaping the climate of Earth's many regions. Surface currents (restricted to the upper 400 m of the ocean) are generally wind-driven and develop their typical clockwise spirals in the northern hemisphere and counter-clockwise rotation in the southern hemisphere (for warm currents). Deep ocean circulation is the result of a number of factors including temperature and salinity variations in water masses, shorelines, subsurface topography, tides, etc.
Salinity	Sea Surface Salinity is a key parameter to estimate the influence of oceans on climate. Along with temperature, salinity is a key factor that determines the density of ocean water and thus determines the convection and re-
Sea level	The sea surface is anything but flat. There are bumps and troughs, all due to different physical characteristics such as gravity, currents, temperature and salinity Since we do not know much about the ocean's bottom, it is easier to refer to "sea height" instead of sea depth. Sea level is measured with reference to a fixed surface height. By analyzing variations from this reference point, scientists determine ocean circulation (currents and eddies at the edges of holes and bumps). seasonal or inter-annual variations, or even longer periods (long-term rise in sea level).
Wind	Surface winds, combined with other atmospheric forces (solar energy, precipitation rate, evaporation rate) are all responsible for the movement of water masses in the ocean, and are thus responsible for ocean currents. Marine winds shape the ocean, and can cause waves as high as a mountain to swell during a storm. They are the source of many legends and color the moods of seafarers around the world
Biogeochemestry	Phytoplankton (vegetable plankton) is the first link in the ocean's food chain, and is the main source of food for most fish. Phytoplankton contains chlorophyll, which instigates photosynthesis in the ocean, absorbs atmospheric CO2 and releases oxygen in sunlight. More than any land-based plant, phytoplankton is the biggest producer of oxygen on Earth

• Marine pressures/activities

3.2 Establishing the context: Defining the stakeholders

In view of the need for consultation in the definition of weights for the criteria, this study proposes the adoption of an approach based on the Best Expert Judgment (BEJ). The BEJ method aims the identification of optimal decisions by providing experts and decision-makers structured framework (Weisberg et al., 2008; Burgman et al., 2011 as cited in Elliott et al., 2018). Despite the uncertainties in complex marine areas, management conclusions can still be reached through integrative assessment (Elliott et al., 2018), experts can synthetize and communicate understanding for societally important questions (Mach et al., 2017). In BEJ, experts can integrate evidence across disciplines, however, they must consider uncertainties and appropriate generalizations across results (Mach et al., 2017).

Bearing in mind the actions developed under the PLASMAR project, the consultation will be applied on the project's partner researchers, once they are currently developing a dense research over the Macaronesia area and, as a result, they also hold the structured framework of PLASMAR project within their research. Within each partner's institution, specialists will be identified for each sector, as observed in Figure 8.

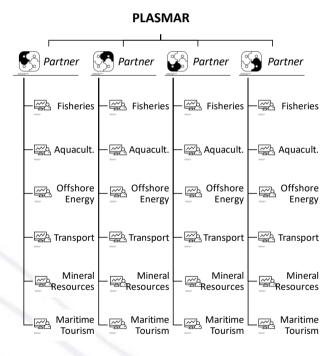


Figure 8: Internal consultation scheme. Image sources: thenounproject.com

Moreover, this consultation aims to be complemented by experts external to PLASMAR project, at the main Universities in Macaronesia or key sectoral experts. Similarly, experts will be selected per sector in each University (Figure 9). This selection can be based on three methods:

- If available, governmental experts' database;
- Snowball sampling, through the indication of experts by the previously consulted PLASMAR experts;

 Internet surveys on scientific platforms for authors of publications related to the sector in the region.

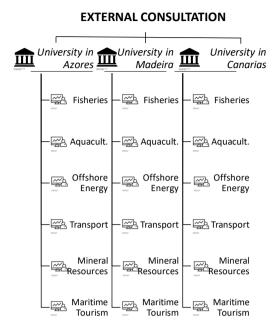


Figure 9: External consultation selection scheme. Image sources: thenounproject.com.

Forman and Peniwati (1998) identify two ways of aggregating individual preferences into a group preference, given the group synergy. The authors observe when individuals are willing to act in concert and pool their judgments in a way that the group becomes a 'new individual', aggregating individual judgments (AIJ) must be used. On the other hand, when individuals are each acting in his/her own right, and aggregation of individual priorities (AIP) must be adopted.

Considering the consultation will take place among experts only and the grades will be given per sector and per region separately, even though the experts will not necessarily act as one single individual once the consultation will be individual, it is assumed initially that the AIJ will be adopted. In this manner, in the aggregation where the reciprocal property is assumed, the geometric mean should be used instead of the arithmetic mean (Forman and Peniwati, 1998; Ossadnik, Schinke and Kaspar, 2016). Nonetheless, the aggregation will be based on a *post hoc* analysis. Moreover, it is important to observe that mathematical aggregation implies in transparency and clarity of results, however, it might not reflect the individual preferences (Ignaccolo *et al.*, 2017).

3.3 Specifying preferences

Considering the benefits previously discussed in the use of AHP methodology, this method will be adopted in the determination of the weight coefficients for the criteria. The methodology will be further explained in steps.

3.3.1 Step 1: Determination of the weight coefficients for the criteria

As previously referred, the AHP method consists on the pairwise comparison of criteria in a matrix and a subsequent comparison of alternatives per criteria in order to output a rank of alternatives (Saaty, 1990). As a result, this method demands the confrontation of the alternatives per criteria, in order to define a final grading.

Considering that in this analysis the alternatives are the maritime sectors defined by PLASMAR as: aquaculture, fisheries, maritime transport, offshore wind energy, aggregate extraction and maritime tourism. Considering also that the confrontation of alternatives – maritime sectors – was not plausible, since the outcome of one specific sector must be seen in the context of this sector and cannot be used for other sectors, and must consider characteristics such as institutional capacity, existent infrastructure for the sector, etc. (Soma, 2003). As a result, the computation of the alternatives scoring was not developed, and the method compared the parameters and clusters for each alternative (maritime sector). The adoption of AHP only in the weight definition had being largely observed in the literature as part of another multicriteria analysis (MCA) method (Babic and Plazibat, 1998; Ragu and Pillai, 1999; Macharis et al., 2004; Bisel Ufuk et al., 2006 apud Konidari and Mavrakis, 2007; Botelho, A.Z. et al. 2015; Soma et al., 2012).

Based on the goal of the PLASMAR action 2.1.2., the identification of areas for blue growth, the pairwise comparison will be performed by cluster-cluster (level 1) per alternative (maritime sector) and later by criteria-criteria (level 2) per alternative (maritime sector). In short, the pairwise comparisons to be performed, based on the cluster (x_1) , criteria (x_2) and alternatives (x_3) , are the following: (x_1) - (x_1) per (x_3) , (x_2) - (x_2) per (x_3) . Moreover, if necessary, sub-criteria – sub-criteria comparisons might be taken.

All the pairwise comparisons should be done for both, contribution and impact (Figure 4). This analysis will be performed through questionnaires to characterize the preferences of the identified experts and later transferred to the pairwise comparison matrixes.

The relative importance for grading the analysis is based on Table 3, scale recommended by Saaty (1990). Fixed point scoring has the advantage of forcing participants to make trade-offs in a decision problem (Hajkowicz *et al.*, 2000). Moreover, Saaty (1990) argues that the AHP is rigorously concerned with the scaling problem and how to correctly combine the priorities resulting from them and, as a result, the scale was maintained in this study.

Table 3: Relative importance scores. Source: Adapted from Saaty (1987).

Intensity of importance	Definition	Explanation		
1	Criteria are equally important	Two criteria contribute equally to the objective		
3	One criterion is slightly more important than the other	Experience and judgment favor one criterion over another		
5	One criterion is more important than the other	Experience and judgment strongly favor one criterion over another		
7	One criterion is strongly more important than the other	A criterion is strongly favored, and its dominance is demonstrated in practice		
9	One criterion is absolutely more important than the other	The evidence favoring one criterion over another is of the highest possible order of affirmation		
2,4,6,8	Intermediate values	Intermediate values		

The different examples of matrices' structures are presented below. In the matrices, each element represents the comparative of the element in the row with the element in the column. For instance, the value a_{12} represents the importance of criteria 1 compared to criteria 2.

Table 4: Pairwise comparison between clusters (x_1) - (x_1)

	1. MSFD	2. MPA	3. Land cover	4. Oceanography	5. Maritime
CLUSTER	GES		& use		Activities
1. MSFD GES	a ₁₁	a ₁₂	a ₁₃	a ₁₄	a ₁₅
2. MPA	a ₂₁	a ₂₂	a ₂₃	a ₂₄	a ₂₅
3. Land cover & use	a ₃₁	a ₃₂	a ₃₃	a ₃₄	a ₃₅
4. Oceanography	a ₄₁	a ₄₂	a ₄₃	a 44	a ₄₅
5. Maritime Activities	a ₅₁	a ₅₂	a ₅₃	a ₅₄	a ₅₅

Table 5: Examples of pairwise comparison between criteria (x_2) - (x_2) within the cluster MPA, for the sectors maritime tourism and aggregate extraction.

MARITIME TOURISM				AGGREGATE EXTRACTION			
	1. Nationally	2. Natura	3. No take		1. Nationally	2. Natura	3. No take
MPA	designated	2000	zone	MPA	designated	2000	zone
	areas				areas		

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1. Nationally	2	2	2	
desig. areas	a ₁₁	a ₁₂	a ₁₃	
2. Natura				
2000	a ₂₁	a ₂₂	a ₂₃	
3. No take	2	2	2	
zone	a ₃₁	a ₃₂	a ₃₃	

1. Nationally				
desig. areas	a ₁₁	a ₁₂	a ₁₃	
2. Natura	2	2	2	
2000	a ₂₁	a ₂₂	a ₂₃	
3. No take	2	2	2	
zone	a ₃₁	a ₃₂	a ₃₃	

3.3.2 Step 2: Calculating the weights per sector

In order to compute the weights for the criteria in AHP from the cluster and criteria matrices, as recommended by Saaty (1990), the normalized values must be derived from each matrix, in which every element in a matrix is divided by the sum of the elements in the column, as observed below.

$$\bar{a}_{ij} = \frac{a_{ij}}{\sum\limits_{k=1}^{n} a_{kj}}$$

Posteriorly, a weight vector is calculated through the average of the normalized elements in each row, through the following equation.

$$w_j = \frac{\sum\limits_{k=1}^n \bar{a}_{ik}}{n}$$

The final output of this step is one vector of weights \mathbf{W} with n elements per maritime sector, as observed below, for cluster and criteria.

$$W = \begin{bmatrix} w_1 \\ w_2 \\ \dots \\ w_n \end{bmatrix} = \begin{bmatrix} w_1 \\ w_2 \\ w_3 \\ w_4 \\ w_5 \end{bmatrix}$$

3.3.3 Step 3: Calculating the consistency ratio (CR)

Consistency implies that if a > b; b > c then a > c (Soma, 2003). In AHP, the consistency ratio is calculated in order to check the consistency of the analysis (for each matrix) and is a result of the consistency index (CI) per random index (RI) and CI / RI < 0.1.

The index is calculated by CI = λ_{max} – n / n – 1, where λ_{max} is the maximum eigenvalue of the matrix and n is the size of the matrix. The λ_{max} must be calculated like so (Saaty, 1977 cited in Franek and Kresta, 2014):

$$\lambda_{\max} = \sum_{j=1}^{n} \frac{(\mathbf{A} \cdot W)_{j}}{n \cdot W_{j}},$$

where \boldsymbol{A} represents the pair-wise comparison matrix and \boldsymbol{W} the eigenvector.

The values for the RI for small problems are shown in Table 6.

Table 6: Values of the Random Index (RI): Source: (Franek and Kresta, 2014)

Size	2	3	4	5	6	7	8	9	10	11	12
RI	0	0.525	0.882	1.110	1.250	1.341	1.404	1.451	1.486	1.514	1.536

In order to avoid inconsistent values, the selection of the critical criteria through the significant parameters was developed to define a clear hierarchy of importance of the different parameters, as well as to reduce the number of criteria to be compared in the pairwise analysis. Moreover, it is important to observe that, in order to achieve a consistent result, the matrix should be reviewed by the respondent in case the CR > 0.1. However, considering that this practice could generate some inflexibility for the respondents and discourage participation, it was not adopted.

3.3.4 Step 4: Calculating the final weight

The final product of this analysis will be two or four matrices for each sector, one for "Contribution" and one for "Impact", and other two for the medium and high significant parameters if analysed separately. These matrices will contain the weights of the critical criteria per sector, as observed in the example below.

Table 7: Example of structure of the final output of the weight analysis

	HIGH	H/MEDIUM RE	ELEVANCE F	ARAMETE	RS			
		CONTRIBUTIO	N/IMPACT	MATRIX				
Cluster Criteria		Maritime Activities weights						
	Aquaculture	Fisheries	M. Transp.	O. W. Energy	Aggregat e E.	Maritime Tourism		
GE	Descriptor 1							

	Descriptor 2		
	Descriptor n		
MPA	Nationally designated		
	Natura 2000 by EEA		
	No take zone		
	CORINE		
cover	Land use		
Land use/cover	Distance to the coast		
Lan	Point and lineal coastal pressures		
	SST mean		
	SBT mean		
	Sea salinity		
арһу	Currents Vel. max		
Oceanography	Wave height Med		
Ö	Wind Vel. Max		
	Salinity Med		
	Mixed layer thickness Med		
	Aquaculture facilities		
	Maritime traffic lanes		
Maritime pressures	Fishery areas		
	Harbours/ports		
	Coastal point pressures (submarine outfalls, gullyies)		
	Aggregate extraction	/-	
		11	

Maritime tourism			
Cables			
Research area			
Military area			
Cruise Tourism			
Coastal Tourism			
Artificial reefs			
Deep sea mining			
Blue biotech			
Seaweed cultivation			
Diving			
Windsurfing			
Wreck			
Ship building			
Dredging / Sand extraction			
Fossil fuel extraction			
Dumping			
Desalination			
Offshore supply/bunkering			

The input values must be the calculated by the multiplication of the cluster weight (W_{x1}) by the correspondent criteria weight vector (W_{x2}) per sector, as presented below:

 $W_{x1} \cdot W_{x2}$ (Per sector)

4 Methodology Remarks

In this report, a methodology is proposed for the weight attribution to the parameters created under the PLASMAR project based on the aggregate extraction analysis. This proposal aimed to adapt the AHP methodology to the project's goal, and, as a result, to achieve a hierarchy definition for the different parameters within each sector for the Macaronesia region. Currently, a pilot questionnaire is being applied in order to test and improve the methodology. Moreover, an excel table based on Goepel (2013) will be provided for the methodology data analysis. Nevertheless, some considerations need to be done about the methodology.

First, this methodology is based on the analysis of the goals/objectives, clusters, criteria and alternatives. Those different analysis structures implied in the limitation of the analysis by the selection and definition given to these structures. In other words, the selection and definition of the aforementioned analysis components over other possibilities have direct impact on the final results, e.g. the incorporation of the environmental component in the criteria instead of the sectors can imply in a sectoral approach in the environmental evaluation.

Second, the application of questionnaires without a workshop to create an open space for trade-offs to define the weights is a limiting factor of this methodology. Moreover, engaging a wider representation of stakeholders and basing group decision on consensus, when possible, results in more democratic and fair results (Ignaccolo *et al.*, 2017).

Finally, the outcomes must be updated along time, with the changes in the sectors and in the parameters studied (Soma, 2003).

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