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24 **Estimation of jellyfish abundance in the south-eastern Spanish coastline by using**  
25 **an explainable artificial intelligence model based on fuzzy logic**

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44

45 **Abstract**

46 Jellyfish swarms have a direct negative impact on human enterprise, specially on  
47 places dependent on the sun and beach economy. The local economy and the health  
48 of bathers may be at risk from the emergence of these gelatinous organisms.  
49 Economic losses can be mitigated by monitoring the occurrence of jellyfish on the  
50 coast. Due to the lack of jellyfish data, environmental citizen science is presented as  
51 an alternative for data collection. In this study, fuzzy logic-based models have been  
52 used to modelize the knowledge from citizen comments collected by the Infomedusa  
53 app. The effect of local climatological factors such as wind speed and direction on  
54 the incidence of jellyfish on the coast was studied. The fuzzy logic-based models  
55 showed that winds perpendicular to the coast lead to a higher occurrence of jellyfish  
56 swarms in central and eastern Malaga, while winds parallel to the coast have a  
57 greater influence in the westernmost coasts. Wind speed has a different effect on  
58 jellyfish incidence depending on the study area and wind direction. Data extracted  
59 from the Infomedusa app can help to address the historical scarcity of scientific data  
60 on jellyfish. This app presents an opportunity for future studies to expand the  
61 knowledge about the occurrence of these organisms on the coasts and may  
62 contribute to the prediction of onshore arrival.

63 **Keywords**

64 Model-ecosystem, Coastal Waters, Explainable Artificial Intelligence, Fuzzy Rules-  
65 Based System, Fuzzy Clustering, Environmental Citizen Science

## 66 **1. Introduction**

67 Since the 1950s, the increase in jellyfish populations has been perceived as a potential  
68 problem (Purcell, 2012). This is because the upwelling of these gelatinous organisms  
69 directly and indirectly interferes with activities such as fishing (Graham et al., 2003;  
70 Knowler, 2005; Nagata et al., 2009; Dong et al., 2010; Boero 2013; Nastav et al., 2013),  
71 aquaculture (Doyle et al., 2008; Delannoy et al., 2011; Bosch-Belmar et al., 2017;  
72 Halsband et al., 2018) or industrial production processes located along the coast  
73 (Daryanabard and Dawson, 2008).

74 Sun and beach tourism is one of the most important **economic** sectors to the Spanish  
75 economy affected by jellyfish blooms (Bernier et al., 2014), particularly in some places  
76 such as the Costa del Sol (Romero-Padilla et al., 2020). This enclave in southern Spain  
77 has undergone territorial, social and economic transformations induced by tourist  
78 activity, which in turn has boosted other economic activities such as fishing, intensive  
79 agriculture, communications and port and commercial activity. In this way, it is  
80 estimated that tourism on the Costa del Sol had an economic impact of 14,442 million  
81 euros in 2019 due to the arrival of a total of 13 million visitors. Jellyfish can  
82 undoubtedly have an impact on tourism, not only because of the negative perception  
83 instilled by specimens stranded on the sand, but also as a consequence of their stings,  
84 which in many cases can cause intense pain when they come into contact with bathers  
85 (De Donno et al., 2014).

86 The economic losses caused by the appearance of jellyfish on beaches can be partially  
87 mitigated by monitoring their populations, statistical analysis of historical records  
88 (Ghermandi et al., 2015) and the development of predictive models that allow the  
89 establishment of control measures and the removal of specimens that reach the beaches

90 (Bellido et al., 2020). However, this faunal group has traditionally been considered an  
91 unimportant component of the marine ecosystem (Brotz et al., 2012) and has been  
92 systematically excluded from the methodological manuals of routine scientific  
93 zooplankton surveys and trawl surveys (Heinle, 1965; Pugh, 1989; Hay, 2006). This  
94 means that historical records on the abundance of jellyfish populations are nowadays  
95 scarce (Brotz et al., 2012).

96 In the absence of scientific data and historical records, crowdsourcing systems through  
97 Environmental Citizen Science (ECS) have emerged as an alternative methodology for  
98 data collection (García-Soto et al., 2021). This is possible due to the the popularisation  
99 of mobile devices, which allow the participation of a large number of citizens in  
100 research and complex scientific problems through sensors and apps integrated in the  
101 smartphones (Boero, 2013). These data are sent to a server where it is possible to  
102 perform data mining, analysis and information dissemination tasks (Zamora et al., 2016;  
103 Komninos, 2018), allowing scientists to overcome limitations of budget, equipment or  
104 spatial scale imposed in most research projects (Zamora et al., 2016).

105 A clear example of non-driven open distributed collaboration, is the monitoring of  
106 relative abundances of jellyfish through the *Infomedusa* app. Infomedusa  
107 (<https://infomedusa.es/>) was developed by the Diputación Provincial de Málaga  
108 (Malaga County Council) in collaboration with the Aula del Mar de Málaga, and has  
109 been used effectively to monitor the presence or absence of jellyfish along the Costa del  
110 Sol (south-eastern Spain). Based on the database generated by this app, it has been  
111 possible to demonstrate: (i) the effectiveness of ECS as a valid pathway for scientific  
112 production from the participation of citizens in science and research, (ii) the effect of  
113 climate oscillations on jellyfish swarms and their incidence on beaches (Bellido et al.,

114 2020), and (iii) the effect of different local climate components on jellyfish abundance  
115 from classical statistical modelling techniques (e.g. **Generalised Additive Models**)  
116 (Gutiérrez-Estrada et al., 2021).

117 In non-driven applications such as Infomedusa, the information generated is not  
118 controlled by a structured scientific programme focused on **obtained conclusions**.  
119 Therefore, the collection of useful data and the obtaining conclusions requires a pre-  
120 processing of the information provided by the users and a subsequent meta-analysis  
121 characterised by the use of non-linearly correlated data associated with a high degree of  
122 uncertainty. Under these circumstances, techniques of explainable artificial intelligence  
123 such as fuzzy logic (FL) show better performance compared to the results obtained with  
124 **others** statistical procedures (Tang and Chi, 2005; Yi et al., 2016, Arrieta et al., 2020).

125 In particular, FL mathematically formalises the way humans interact with the  
126 environment and make decisions. Thus, models based on FL allow dealing with  
127 imprecise information in terms of fuzzy or fuzzy sets, making it possible to model any  
128 non-linear process and learn from the data through different learning algorithms (Martín  
129 del Brío and Sanz Molina 2006). In this way, FL models are a very good alternative for  
130 data processing in an environmental citizen science context. The bases of the proposed  
131 methodology (Aroba, 2003) have demonstrated its efficacy for modelling the qualitative  
132 behaviour of complex systems in diverse contexts, like: environmental (Fernández-  
133 Camacho et al., 2015), biochemical (Luis et al., 2018) or medical (Enríquez et al.,  
134 2018), where the obtained results were validated by experts in the corresponding fields,  
135 confirming the veracity and value of the generated qualitative information.

136 Therefore, this paper presents a methodology based on a FL approach for the  
137 characterization of the effect of local environmental conditions (wind direction and

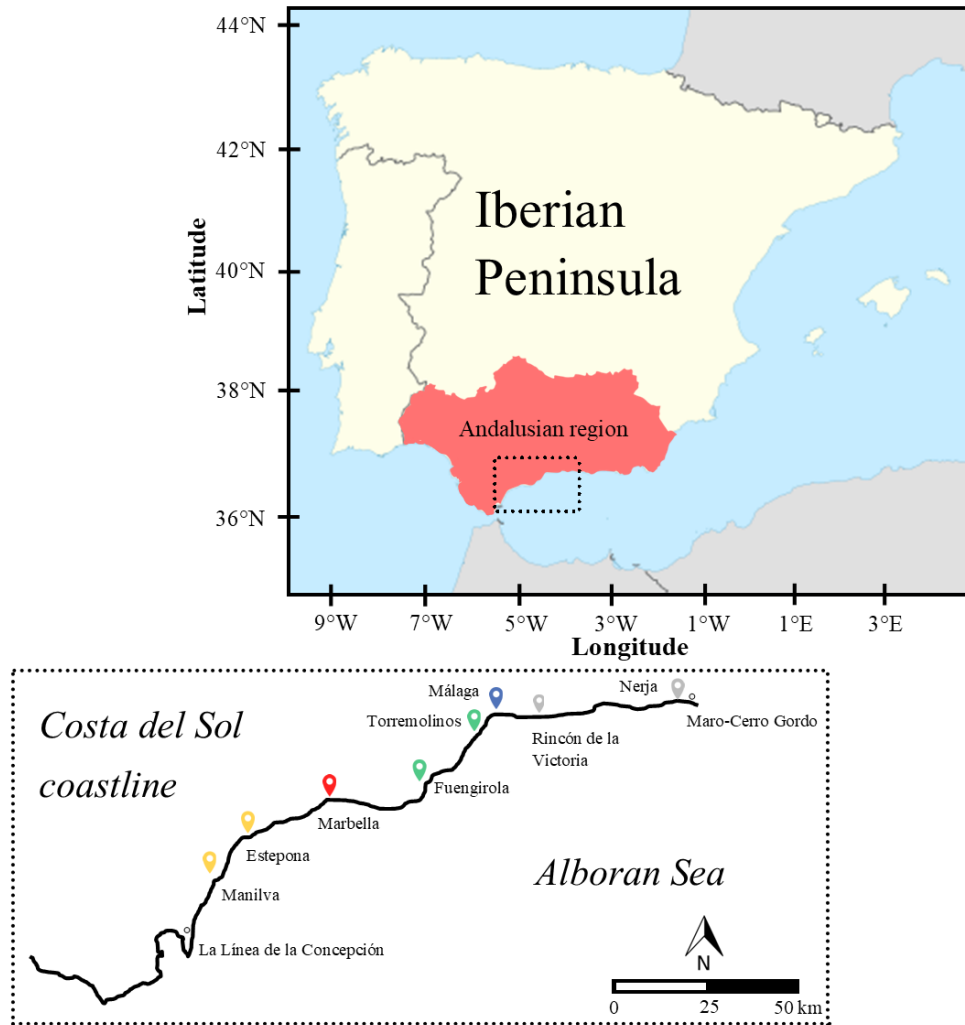
138 wind speed) on the incidence of jellyfish on the beaches of the Costa del Sol, located in  
139 the Andalusian coastline (southern Spain).

140

## 141 **2. Material and methods**

### 142 **2.1. Study area**

143 The study area is located in the province of Malaga (Andalusian region, south-eastern  
144 Spain). The Costa del Sol stretches from La Línea de la Concepción on the border with  
145 the province of Cádiz in the west to the cliffs of Maro-Cerro Gordo on the border with  
146 the province of Granada in the east, for approximately 185 km (Fig. 1). The Costa del  
147 Sol is made up of three distinct areas: (1) Western Costa del Sol, specialising in foreign  
148 tourism; (2) Eastern Costa del Sol, oriented towards inland and residential tourism; and  
149 (3) the capital area, with administrative, commercial and non-tourist residential  
150 functions (García and Ocaña, 1982).



151

152 Figure 1. Study area. Location of the main beach sets cities. Colours on the mark point represent  
 153 membership to the different beach sets: Set 1 (Blue), Málaga beaches; Set 2 (Gray), Nerja-Rincón de la  
 154 Victoria beaches; Set 3 (Green), Torremolinos-Fuengirola beaches; Set 4 (Red), Marbella beaches; Set 5  
 155 (Yellow), Manilva-Estepona beaches. The border localities of the Costa del Sol to the west (La Línea de  
 156 la Concepción) and to the east (Maro-Cerro Gordo) are also marked.

157

158 The Alboran Sea bathes the coastline of the study area and its surface circulation is  
 159 characterised by the presence of two medium-sized anticyclonic gyres (Western and  
 160 Eastern Anticyclonic Gyre) surrounded by the meandering flow of Atlantic surface

161 water, the Atlantic Jet. Both anticyclonic gyres are separated by a cyclonic structure of  
162 more elusive existence called the Central Cyclonic Gyre (Peliz et al., 2013; Macías et  
163 al., 2016; García-Lafuente et al., 2021).

164 A total of 149 beaches belonging to 14 different localities have been analysed in this  
165 study, from Ancha de Casares beach in the west to Las Alberquillas beach in the east  
166 (Fig. 1).

## 167 ***2.2. Jellyfish relative abundance data***

168 The relative abundance of jellyfish was obtained from the Infomedusa app, available in  
169 this google link: <https://play.google.com/store/apps/details?id=es.infomedusa&hl=es>.  
170 Infomedusa is a citizen science app available for iOS and Android smartphones and  
171 developed by the organisation Aula del Mar de Málaga (<https://www.auladelmar.info/>).  
172 Users have real-time knowledge of climatological and oceanographic variables through  
173 the app. The application has a functionality that allows users to chat in an individual  
174 forum for each beach about climate conditions and the presence or absence of jellyfish.  
175 A database is generated from the citizens' comments. In 2018 (between 25 January and  
176 3 October), 40,276 comments were registered, of which 9,963 were validated and  
177 selected for the purpose of this study.

178 Since comments from citizens had a natural language structure, a transition network  
179 (TN) was designed for their processing. TNs has been widely used in different fields of  
180 knowledge (Woods, 1973; De Carolis et al., 1996; Gangopadhyay, 2001; Gutiérrez-  
181 Estrada et al., 2005). These systems are based on a data structure consisting of a set of  
182 nodes representing finite states which are connected through a series of arcs. This model  
183 consists on a transition between states according to the grammatical category of the

184 word read (Woods, 1970). The state changes are determined by the grammatical  
185 category of the word read, with the input being the user's comment in the chat and the  
186 final output being the information corresponding to the number of jellyfish. The outputs  
187 of the system were set to a series of discrete values: 'None' (0); 'Few' (1), 'Some' (2),  
188 'Many/Much' (3), and 'Very many/very much' (4). A detailed explanation of this  
189 process can be found in Gutiérrez-Estrada et al. (2021).

190

### 191 ***2.3. Environmental data***

192 Due to the planktonic nature of this species, climate could allow huge accumulations of  
193 jellyfish along the coastline. Two of the main factors influencing jellyfish strandings are  
194 the direction and speed of the prevailing local winds that cause the migration of surface  
195 water towards the coast (Graham et al., 2001; Bellido et al., 2020; García-Lafuente et  
196 al., 2021; Gutiérrez-Estrada et al., 2021). Winds in the study area are characterised by  
197 seasonal behaviour. In winter-spring they have a north-westerly component and in  
198 summer the south-easterly component predominates (Mercado et al., 2012).

199 Wind data was obtained from the Spanish Meteorological Agency  
200 (<http://www.aemet.es/en/portada>). Five different set of beach localities and four  
201 different weather stations were used to deal with differences in coastlines morphology  
202 and orientation (Table 1).

203

204

205

206 Table 1. Description of the weather stations and localities covered.

<b>Weather station</b>	<b>Location</b>	<b>Altitude (m)</b>	<b>Set of beach localities*</b>	<b>Localities covered (From East to West)</b>
Málaga	36° 39'N, 4° 28'W	5	1, 2	Nerja, Torrox, Algarrobo, Vélez-Málaga, Rincón de la Victoria, Málaga
Fuengirola	36° 32'N, 4° 37'W	8	3	Torremolinos, Benalmádena, Mijas, Fuengirola
Marbella	36° 29'N, 4° 44'W	2	4	Marbella
Manilva	36° 22'N, 5° 15'W	140	5	Estepona, Casares, Manilva

207 \*See Gutiérrez-Estrada et al. (2021)

208

209 These stations record hourly data from temperature (average, maximum and minimum,  
 210 in °C), wind (gust, mean and maximum speed, and direction, in km/h), atmospheric  
 211 pressure (hPa), precipitations (mm) and humidity (%). For the purpose of this study,  
 212 only mean speed and direction were recorded. Data for each variable were daily  
 213 averaged.

214

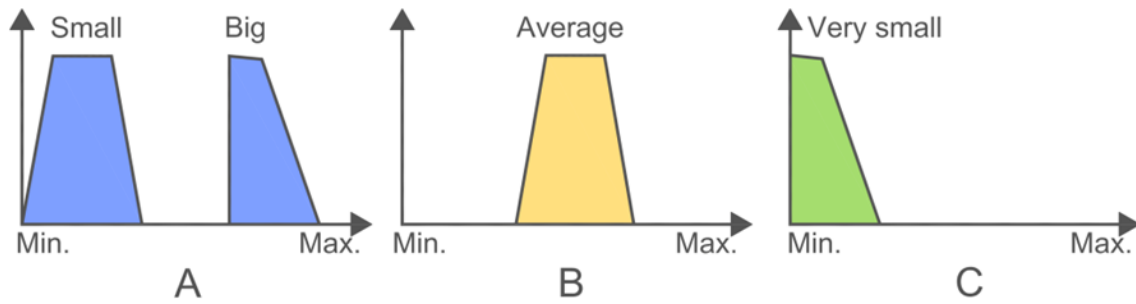
#### 215 **2.4. Proposed Fuzzy Logic Approach**

216 Currently, Artificial Intelligence is reaching high levels of applicability. Within  
 217 Artificial Intelligence techniques, those that allow us to understand the knowledge they  
 218 have learned from the data are of greatest interest particularly in those areas where the  
 219 comprehension of the models is important. This understanding is of vital importance to  
 220 be able to certify its operation according to our trustworthy standards. This highly  
 221 topical field is currently known as eXplainable Artificial Intelligence (known as XAI)

222 (Barredo et al., 2020). Inside the XAI techniques is FL (Fernández et al., 2019), which  
223 is one of the most successful areas of Artificial Intelligence in applied problems due to  
224 its expressive richness using linguistic terms to enunciate real world concepts and  
225 definitions. In this work we propose the use of a fuzzy rules-based model obtained  
226 through clustering mechanisms, which is described below.

227 The proposed FL approach (Aroba, 2003) is based on the methodology described in  
228 Sugeno et al. (1993), to build fuzzy models based on the fuzzy C-Means algorithm  
229 (Bezdek, 1981; Höppner and Klawonn, 2003) to generate fuzzy rules with this format:  
230 IF  $x \in A$  THEN  $y \in B$ , where  $x = (x_1, x_2, \dots, x_n) \in \mathfrak{R}^n$  are input variables  
231 (antecedents),  $A$  are  $n$  fuzzy sets,  $y \in \mathfrak{R}$  is the output variable (consequent) and  $B$  is a  
232 fuzzy set for this variable.

233 The first step, before data processing, **is to determine** the variables of the dataset whose  
234 behaviour needs to be studied, so that, these selected variables will be the consequents  
235 in the fuzzy rules that will be generated. Therefore, the rest of parameters of the dataset  
236 will play the role of antecedents in the fuzzy rules. Then, the proposed methodology  
237 analyses automatically the dataset, so that the consequent parameters are fuzzy clustered  
238 in an optimum number of fuzzy clusters are obtained (Fukuyama and Sugeno, 1989).  
239 Once the consequent parameters have been clustered, each fuzzy cluster is projected  
240 onto the antecedent space (Sugeno et al., 1993), obtaining the membership grade of the  
241 antecedents to each fuzzy cluster. Finally, with the obtained information in the previous  
242 steps, is generated a set of graphical fuzzy rules, like in the Figure 2, where the  
243 antecedent variables are  $A$  and  $B$ , and the consequent variable is  $C$ .



244

245 Figure 2. Example of graphical IF–Then fuzzy rule.

246 The fuzzy rule of the Figure 2 can be interpreted in natural language as:

247 IF  $A$  is small or big AND  $B$  is average, THEN  $C$  is very small.

248 The proposed fuzzy approach has been designed to be applied on multi-parametric  
 249 quantitative datasets to generate automatically a Fuzzy Rules-Based System (FRBS)  
 250 that provides a qualitative model of the processed data.

251

## 252 **2.5. Comparison between the proposed FRBS models and GAMs**

253 In order to compare the results obtained with FRBS models and Generalised Additive  
 254 Models (GAMs), different values of interest or hot-points were inferred from the fuzzy  
 255 rules generated for the antecedents (explanatory variables) wind direction and wind  
 256 speed. Hot-points were established according to the behaviour of the partial residuals of  
 257 generalised additive models (GAMs) calibrated by Gutiérrez-Estrada et al. (2021) with  
 258 the variation of the explanatory variables. The hot-points values were fixed into the  
 259 FRBS and the results were compared with the GAM estimations.

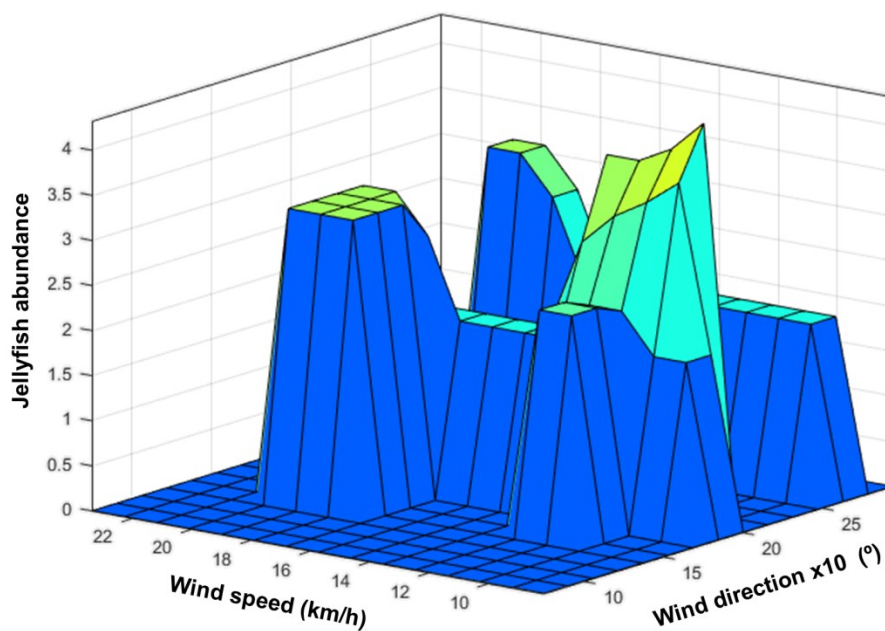
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262 **3. Results**

263 For a better comprehension of the fuzzy rules obtained in each analysed zone, the  
264 results have been exported to the format of the Fuzzy Logic Toolbox of Matlab  
265 (MathWorks®), with the goal of representing the rules in 3D surface format, so that be  
266 more readable and interpretable in natural language.

267 For the Set 1 of beach localities (Malaga city beaches), a total of 7 fuzzy rules were  
268 generated by the fuzzy logic-based methodology. The 3D surface representation of  
269 these 7 fuzzy rules is presented in Figure 3, where relative jellyfish abundance act as  
270 consequent. The highest incidence of jellyfish is shown when the wind direction is  
271 around 225° and the speed is 10.5 km h<sup>-1</sup>. The incidence of jellyfish is null with a wind  
272 direction of 135° and 230-250° or a wind speed higher than 21 km h<sup>-1</sup>. The results of the  
273 model when setting the input variables at the established hot-points are represented in  
274 Table 2.


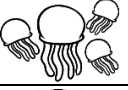









275

276 Figure 3. Surface view results of the Set 1 FRBS.

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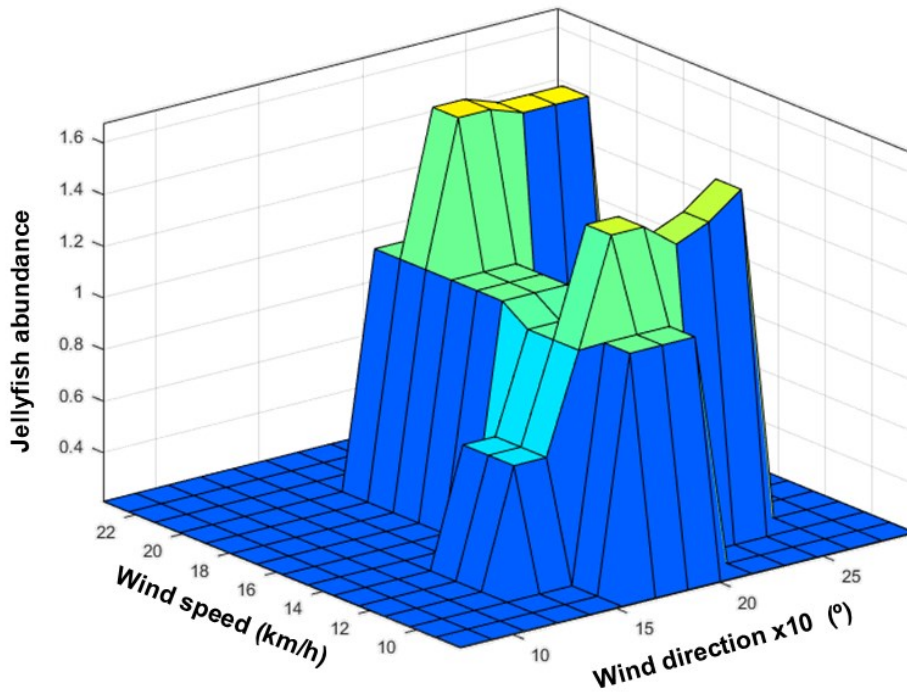
278 Table 2. Schematic results of the jellyfish incidence in Set 1 after setting the input variables in the  
 279 Gutiérrez-Estrada et al. (2021) hotspots.

Wind speed hotspots (km/h)	Wind direction hotspots (° x10)		
	18	21	25
10.5			
14			
19			

280

281 The results of fuzzy logic-based model show that a decrease of jellyfish incidence is  
 282 driven by values of wind direction and wind speed around 250° or 10 km h<sup>-1</sup>,  
 283 respectively. However, the level of explained variance by the fuzzy approach is clearly  
 284 lower than the reached by GAMs calibrated by Gutiérrez-Estrada et al. (2021) (GAM  
 285  $r^2=0.77$ ;  $Fuzzy_{presence} r^2=0.20, p<0.05$ ).

286 For the Set 2 of beach localities (beaches from Nerja to Rincón de la Victoria), a total of  
 287 6 fuzzy rules were generated by the fuzzy logic-based methodology. The 3D surface  
 288 representation of the resulting fuzzy rules (Fig. 4) show a relatively low presence of  
 289 jellyfish for the whole study period, although the highest presence was concentrated in  
 290 winds from the south (180-200°). Jellyfish only appeared at wind speeds between 9 and  
 291 21 km h<sup>-1</sup>. Similar to the results of Gutiérrez-Estrada et al. (2021), wind directions  
 292 above 230° lead to a decrease of jellyfish incidence (Table 3). Again, GAMs for the Set  
 293 2 provided better fits versus FRBS (GAM  $r^2=0.75$ ;  $Fuzzy_{presence} r^2=0.24, p<0.05$ ).












294

295 Figure 4. Surface view results of the Set 2 FRBS.

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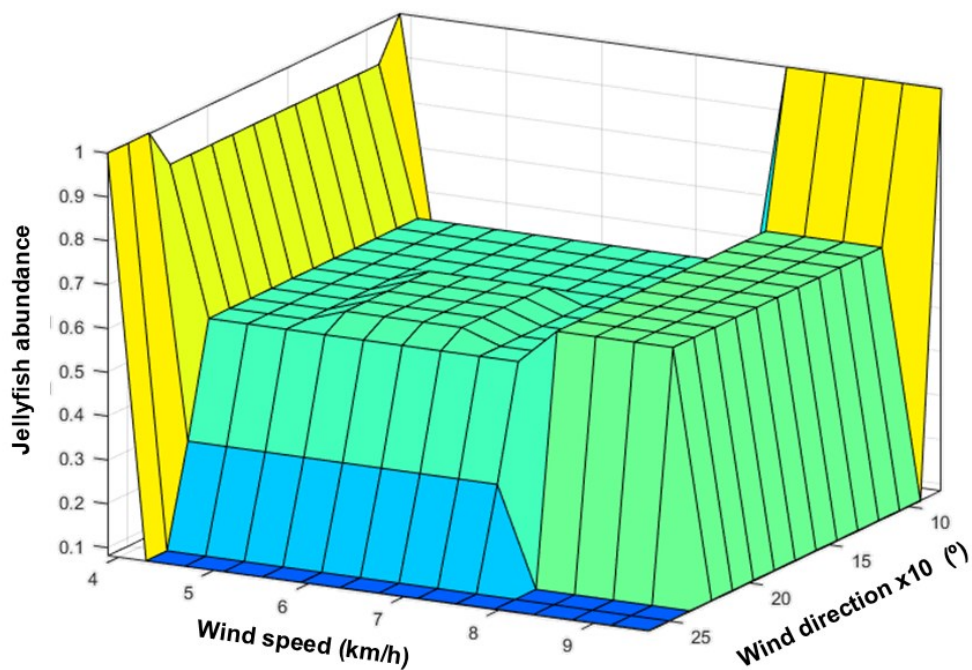
297 Table 3. Schematic results of the jellyfish incidence in Set 2 after setting the input variables in the  
 298 Gutiérrez-Estrada et al. (2021) hotspots.

Wind speed hotspots (km/h)	Wind direction hotspots (° x10)		
	17.5	21.5	26
10			
13			
18			

299

300 For the Set 3 of beach localities (beaches from Torremolinos to Fuengirola), a total of 4  
 301 fuzzy rules were generated by the fuzzy logic-based methodology. In this case, GAMs  
 302 and FRBS provided the worst fit capacity which was reflexed in a low level of

303 explained variance (GAM  $r^2=0.37$ ; Fuzzy<sub>presence</sub>  $r^2=0.12$ ,  $p<0.05$ ). The 3D surface  
304 representation of the obtained fuzzy rules is shown in Figure 5. The environmental  
305 variables analysed remained relatively constant for the whole study period, which  
306 makes it difficult to analyse the effect on jellyfish abundance. The recorded wind speed  
307 did not exceed  $10 \text{ km h}^{-1}$  as in the case of the previous areas, so the effect of this  
308 variable cannot be interpreted in a precise way. However, according to Gutiérrez-  
309 Estrada et al. (2021), the results of the FRBS show how the incidence of jellyfish starts  
310 to decrease when the wind direction is higher than  $225^\circ$  (Table 4).



311
















312 Figure 5. Surface view results of the Set 3 FRBS.

313

314

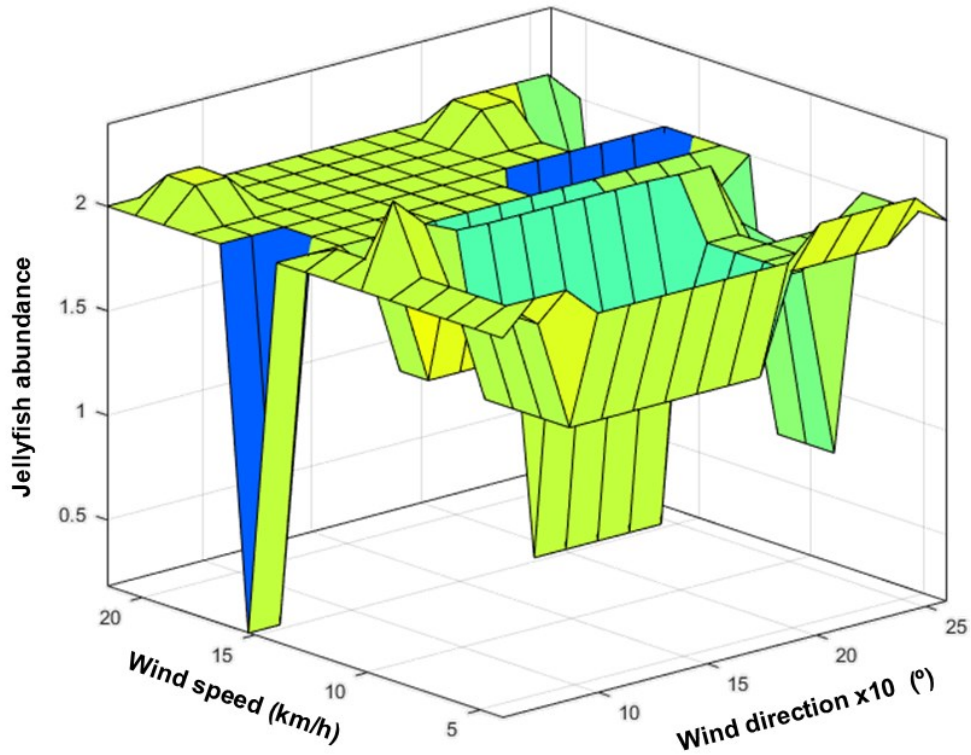
315

316 Table 4. Schematic results of the jellyfish incidence in Set 3 after setting the input variables in the  
 317 Gutiérrez-Estrada et al. (2021) hotspots.

Wind speed hotspots (km/h)	Wind direction hotspots ( $^{\circ}$ x10)		
	17	21	22.5
5			
5.5			
6			
6.5			
7			

318

319 For the Set 4 of beach localities (Marbella beaches), a total of 12 fuzzy rules were  
 320 generated by the fuzzy logic-based methodology which implied a significant  
 321 improvement respect to the Sets 1, 2 and 3 and provided better results than GAMs  
 322 (GAM  $r^2=0.41$ ; Fuzzy<sub>presence</sub>  $r^2=0.53$ ,  $p<0.05$ ). The 3D surface representation of the  
 323 FRBS is shown in Figure 6. The results show that the lowest number of jellyfishes  
 324 events are recorded when the wind speed is set at around 15 km h<sup>-1</sup> (medium-high in the  
 325 time series analysed) with a wind direction of 180°. The highest incidence of jellyfish is  
 326 recorded with a wind direction of 90° and a wind speed of 10.4 km h<sup>-1</sup>. The incidence of  
 327 jellyfish remained at a medium value (around 2, on a scale 0-5) for all cases of wind  
 328 directions and speeds. Results of the FRBS with the established hot-points are shown in  
 329 Table 5.


















330

331 Figure 6. Surface view results of the Set 4 FRBS.

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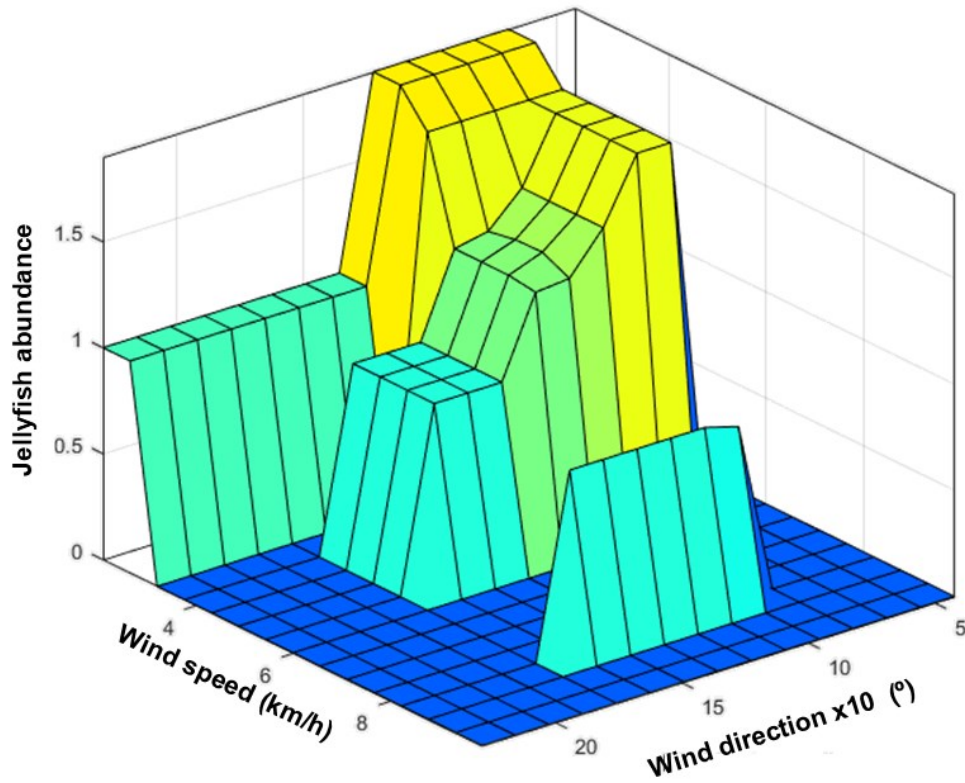
333 Table 5. Schematic results of the jellyfish incidence in Set 4 after setting the input variables in the  
 334 Gutiérrez-Estrada et al. (2021) hotspots.

Wind speed hotspots (km/h)	Wind direction hotspots (° x10)		
	8	20	24
7.5			
10.5			
12.5			
14.5			
16			

335

336 According to Gutiérrez-Estrada et al. (2021), wind directions higher than  $200^\circ$  tend to  
337 have a reducing effect on the incidence of jellyfish at certain wind speeds. The results  
338 for the wind speed set at  $16 \text{ km h}^{-1}$  are a particular case of null jellyfish incidence  
339 between  $14$  and  $16 \text{ km h}^{-1}$  at wind directions of less than  $80^\circ$  and  $180$ - $250^\circ$  that has been  
340 generated by the FRBS.

341 Finally, the 6 fuzzy rules that were generated by the fuzzy logic-based methodology for  
342 the Set 5 of beach localities (beaches from Manilva to Estepona), were not enough to be  
343 able to explain more than 32% of variance. The 3D surface representation of the FRBS  
344 is shown in Figure 7. The results showed that jellyfish incidence is reduced by high  
345 wind speeds and is increased by wind directions between  $70$  and  $140^\circ$ . However, a  
346 specific case is presented in which the incidence of jellyfish may be favoured in this  
347 same range of directions if the wind speed is sufficiently intense. The results for this  
348 area are in agreement with Gutiérrez-Estrada et al. (2021), whose results show the  
349 existence of the same range of directions in which the number of jellyfish increases, as  
350 well as that high wind intensity decreases the jellyfish incidence (Table 6).



351

352 Figure 7. Surface view results of the Set 5 FRBS.

353

354 Table 6. Schematic results of the jellyfish incidence in Set 5 after setting the input variables in the  
 355 Gutiérrez-Estrada et al. (2021) hotspots.

Wind speed hotspots (km/h)	Wind direction hotspots (°x10)				
	6	9.5	11.5	13.5	17
4.5					
5.5					
7.5					

356

357

358

#### 359 **4. Discussion**

360 In this paper, a set of fuzzy logic-based models have been calibrated as an alternative to  
361 traditional statistic approaches for the analyses and interpretation of citizen science  
362 information on stranding of jellyfishes on Costa del Sol beaches. The proposed heuristic  
363 approach has been selected because FL allows to take intermediate membership values  
364 between data sets from information extracted from natural language reasoning, thus  
365 providing formal support for imprecise information (Martín del Brio and Sanz-Molina,  
366 2006). This feature of FL is clearly very interesting from an ecological point of view  
367 because it **allows to find** non-linear relationships in multidimensional systems in a  
368 relatively easy way. This explains the increase in the last decade of applications of this  
369 heuristic technique to ecological applications of FRBS on ecology problems (Albajes-  
370 Eizagirre et al., 2011; Brotz et al., 2012; Gutiérrez-Estrada et al., 2013; Colbach et al.,  
371 2020; Balbuena et al., 2021).

372 On the other hand, the development of this study takes place in one of the most touristic  
373 places in Spain, which makes the study area a place of special interest for testing non-  
374 driven sources of information such as citizen science. Part of the time series of citizen  
375 comments analysed takes place in summer, which coincides with the peak of tourists in  
376 the area and with the period of highest incidence of jellyfish swarms (Hamner and  
377 Dawson, 2009) which supposes an opportunity to address the scarcity of temporal data  
378 on these gelatinous organisms.

379 The obtained FRBS have qualitatively agreed with the models calibrated by Gutiérrez-  
380 Estrada et al. (2021) although in 4 out of 5 groups of beaches, the explained variance

381 levels were lower than those obtained by GAMs. This is in line with other works in  
382 which these approaches have been compared. For example, Gutiérrez-Estrada et al.  
383 (2012) modelling inflow rates for the water exchange management in semi-intensive  
384 aquaculture ponds in the south of Spain found that GAMs and fuzzy logic-based models  
385 provided similar response functions and different levels of explained variance. Also,  
386 Gutiérrez-Estrada et al. (2013) analysing the consistency of fuzzy rules to predict the  
387 water beetle diversity in temporary and fluctuating waters indicated that GAMs and  
388 FRBS had similar behaviours. However, these authors highlighted an important  
389 advantage of fuzzy approaches versus GAMs respect to the interpretation of the results.  
390 While in the case of GAMs the analysis of how the predictors are related to the  
391 dependent variable is carried out computing the partial residuals (which allow to  
392 evaluate the nature of the relationship between the predictor with the adjusted  
393 dependent variable value and hence the nature of the influence of the respective  
394 predictor in the overall model), in the FRBS the results are interpreted as rules  
395 associated to different linguistic categories which is very close to how humans solve  
396 everyday problems.

397 Given the planktonic nature of jellyfishes, the occurrence of these organisms swarms is  
398 increased or decreased by large-scale climatological factors (Bellido et al., 2020) and  
399 mainly can be driven by local climatological and oceanographical factors such as wind  
400 and sea currents (Zavodnik, 1987; Canepa et al., 2014). *Pelagia noctiluca* is the most  
401 common and conspicuous jellyfish species in the Mediterranean Sea (Canepa et al.,  
402 2014). According to citizen's comments, this species was the most frequent on the coast  
403 of Malaga and **the only one species** which appeared in high concentrations **during the**  
404 **analysed time range**. This species passive transport towards the coast is closely related  
405 to horizontal currents. Consequently, the spatio-temporal interpretation of their

406 occurrences arises from the interaction between population dynamics and their dispersal  
407 by currents (Hamner and Dawson, 2009; Berline et al., 2013).

408 The Mediterranean thermohaline circulation is the main driven of the Alboran Sea  
409 currents and the Atlantic Jet dominates the circulation in the first 200 m. When the  
410 Atlantic Jet flows through the Strait of Gibraltar and passes close to the Spanish coast, it  
411 generates two cyclonic circulation cells, one coinciding with this study the sets of  
412 beaches 1-3, and the other one with the sets of beaches 4 and 5 (Vargas-Yáñez et al.,  
413 2021). These cyclonic cells produce divergence of surface waters at the inner part of the  
414 gyres, resulting the upwelling of deeper, colder, and nutrient-rich waters on the coast  
415 (Ramírez et al., 2021). But this surface circulation is also subject to seasonal changes.  
416 This way, in our study area westerly winds prevail during the spring and easterlies of  
417 lesser intensity prevails during summer (Vargas-Yáñez et al., 2021). The summer wind  
418 condition strongly influences on the stranding of jellyfishes on Costa del Sol beaches  
419 and significantly increase the probability that a user of Infomedusa records a jellyfish  
420 sighting. This is because the coast line will be perpendicular to the resultant force of  
421 summer east wind and the currents generated by the two cyclonic circulation cells. The  
422 probability of jellyfish stranding clearly will decrease in winter and spring because the  
423 prevailing winds will force the jellyfish to move offshore.

424 This general behaviour not fulfilled for all sets of beaches. For example, for the set of  
425 beaches 4 and 5 (western beaches of study area), our results indicate that the incidence  
426 of jellyfish is increased when the winds are parallel to the coast. These results confirm  
427 the findings of Bellido et al. (2020), who suggested that in the western Costa del Sol an  
428 accumulation of jellyfish occurs due to the Western Anticyclonic Gyre and then  
429 jellyfish are transported eastwards by lateral winds. These authors also determined that

430 the climatic indices North Atlantic Oscillation (NAO) and Arctic Oscillation (AO) have  
431 an effect on the incidence of jellyfish swarms by influencing their reproduction and  
432 primary production in the area. Jellyfish swarms have a different response to changes in  
433 temperature depending on the area: in colder areas, it is the extreme cold events that  
434 promote these swarms, while in warmer areas such as our study area, **there are the**  
435 **warmer events** (Purcell, 2005; Bellido et al., 2020). The effect of high temperature  
436 events has been shown to favour the appearance of different jellyfish species in the  
437 Atlantic margin of the Iberian Peninsula (Rodríguez-García et al., 2021), so this variable  
438 should be taken into account in future work.

439 Prediction of these swarming events can be elusive due to the evasive behaviour of  
440 these species, which are able to swim upstream or dive to avoid them (Zavodnik, 1987;  
441 Malej, 1989; Fossette et al., 2015; Bellido et al., 2020). The occurrence of swarms of  
442 this jellyfish species has been shown to have a negative effect on small pelagic fisheries  
443 in the Spanish Mediterranean (Baez et al., 2021). Moreover, *Pelagia noctiluca* stings  
444 have been the cause of most cases of jellyfish stings on the Malaga coast (Canepa et al.,  
445 2014; Templado et al., 2021), so the use of intelligent models such as FL systems for  
446 improve the management of jellyfish swarm events are very useful tools for both the  
447 health of bathers and the economy of coastal touristic sites (Gutiérrez-Estrada et al.,  
448 2021). Nevertheless, jellyfish are key component for marine ecosystems, which implies  
449 that should be managed as organisms delivering both risks and benefits to society  
450 (Hamner and Dawson, 2009; Duarte et al., 2014). For this purpose, citizen science could  
451 have an important role to complement and enhance scientific studies and governments  
452 in actions for which they have not enough resources (Lee et al., 2006).

## 453 **5. Conclusions**

454 The results of the study developed, show that the use of artificial intelligence techniques  
455 such as FL, can address the scarcity of data on the incidence of jellyfish on the coast.  
456 The obtained fuzzy rules-based systems have made it possible to modelize the  
457 knowledge of citizens' natural language comments in the Infomedusa app. It has also  
458 been confirmed that local environmental variables such as wind direction and wind  
459 speed play a key role in the occurrence of jellyfish swarms in the province of Malaga.  
460 For the central and easternmost areas of the province, winds perpendicular to the coast  
461 lead to a higher occurrence of jellyfish swarms. However, for the westernmost areas the  
462 highest occurrence happens with winds parallel to the coast. Depending on the area and  
463 wind direction, high wind speed can favour or diminish the appearance of jellyfish.

464 From this study as a starting point, the application of FRBS and their use with the data  
465 obtained from the Infomedusa app, allows to use citizen collaboration to understand the  
466 climatic factors which drive the upwelling of these gelatinous organisms on the coasts.  
467 Future studies would allow to address the socio-economic impact of jellyfish on areas  
468 economically dependent on sun and beach tourism.

469

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#### 475 **Conflict of interest**

476 The authors have no conflicts of interest to declare.

477

478 **Author contributions**

479 **Castro-Gutiérrez, J., Gutiérrez-Estrada J.C., Aroba, J.:** Conceptualization,  
480 Methodology, Software, Validation, Formal Analysis; Data Curation, Writing-Original  
481 Draft; **Pulido-Calvo, I., Peregrín, A.:** Conceptualization, Methodology, Formal  
482 Analysis, Writing-Review and Editing; **Báez, J.C.:** Investigation; Writing-Review and  
483 Editing; **Bellido, J.J., Souviron-Priego, L.:** Investigation, Resources.

484

485 **References**

- 486 Albajes-Eizagirre, A., Romero, L., Soria-Frisch, A., & Vanhellefont, Q.  
487 (2011). Jellyfish prediction of occurrence from remote sensing data and a non-  
488 linear pattern recognition approach. *Remote Sensing for Agriculture,*  
489 *Ecosystems, and Hydrology XIII*. <https://doi.org/10.1117/12.898162>
- 490 Aroba, J. (2003). *Advances in the decision making in software development projects*.  
491 PhD Thesis, University of Sevilla, Spain.
- 492 Arrieta, A. B., Díaz-Rodríguez, N., Del Ser, J., Bennetot, A., Tabik, S., Barbado, A.,  
493 García, S., Gil-López, S., Molina, D., Benjamins, R., Chatila, R., & Herrera, F.  
494 (2020). Explainable Artificial Intelligence (XAI): Concepts, taxonomies,  
495 opportunities and challenges toward responsible AI. *Information fusion* 58: 82-  
496 115. <https://doi.org/10.1016/j.inffus.2019.12.012>
- 497 Báez, J. C., Pennino, M. G., Albo-Puigserver, M., Coll, M., Giraldez, A., & Bellido, J.  
498 M. (2021). Effects of environmental conditions and jellyfish blooms on small  
499 pelagic fish and fisheries from the Western Mediterranean Sea. *Estuarine,*  
500 *Coastal and Shelf Science* 264: 107699.  
501 <https://doi.org/10.1016/j.ecss.2021.107699>

- 502 Balbuena, J. A., Monlleo-Borrull, C., Llopis-Belenguer, C., Blasco-Costa, I., Sarabeev,  
503 V. L., & Morand, S. (2021). Fuzzy quantification of common and rare species in  
504 ecological communities (FuzzyQ). *Methods in Ecology and Evolution* 12(6):  
505 1070-1079. <https://doi.org/10.1111/2041-210X.13588>
- 506 Bellido, J. J., Baez, J. C., Souviron-Priego, L., Ferri-Yáñez, F., Salas, C., López, J. A.,  
507 & Real, R. (2020). Atmospheric indices allow anticipating the incidence of  
508 jellyfish coastal swarms. *Mediterranean Marine Science* 21(2): 289-297.  
509 <https://doi.org/10.12681/mms.20983>
- 510 Berline, L., Zakardjian, B., Molcard, A., Ourmieres, Y., & Guihou, K. (2013).  
511 Modeling jellyfish *Pelagia noctiluca* transport and stranding in the Ligurian  
512 Sea. *Marine pollution bulletin* 70(1-2): 90-99.  
513 <https://doi.org/10.1016/j.marpolbul.2013.02.016>
- 514 Bernier, E. T., Sánchez, R. R., & Díaz, B. R. (2014). La crisis económica en el sector  
515 turístico. Un análisis de sus efectos en la Costa del Sol. *Revista de análisis*  
516 *Turístico* 18: 11-18.
- 517 Bezdek, J. C. (1981). *Pattern Recognition with Fuzzy Objective Function Algorithms*.  
518 Plenum Press.
- 519 Boero, F. (2013). Review of jellyfish blooms in the Mediterranean and Black Sea.  
520 *Studies and Reviews. General Fisheries Commission for the Mediterranean*, 93.
- 521 Bosch-Belmar, M., Azzurro, E., Pulis, K., Milisenda, G., Fuentes, V., Yahia, O. K. D.,  
522 Micallef, A., Deidun, A., & Piraino, S. (2017). Jellyfish blooms perception in  
523 Mediterranean finfish aquaculture. *Marine Policy* 76: 1-7.  
524 <https://doi.org/10.1016/j.marpol.2016.11.005>
- 525 Brotz, L., Cheung, W. W. L., Kleisner, K., Pakhomov, E., & Pauly, D.  
526 (2012). Increasing jellyfish populations: trends in Large Marine Ecosystems.  
527 *Hydrobiologia* 690(1): 3–20. <https://doi.org/10.1007/s10750-012-1039-7>
- 528 Canepa, A., Fuentes, V., Sabatés, A., Piraino, S., Boero, F., Gili, J. M. (2014). *Pelagia*  
529 *noctiluca* in the Mediterranean Sea. In K. A. Pitt & C. H. Lucas (Eds.), *Jellyfish*  
530 *blooms* (pp. 237-266). Springer Science+Business Media.

- 531 Colbach, N., Chauvel, B., Messean, A., Villerd, J., & Bockstaller, C. (2020). Feeding  
532 pollinators from weeds could promote pollen allergy. A simulation  
533 study. *Ecological Indicators* 117: 106635.  
534 <https://doi.org/10.1016/j.ecolind.2020.106635>
- 535 Daryanabard, R., & Dawson, M. N. (2008). Jellyfish blooms: *Crambionella orsini*  
536 (Scyphozoa: Rhizostomeae) in the Gulf of Oman, Iran, 2002–2003. *Journal of*  
537 *the Marine Biological Association of the United Kingdom* 88(3): 477-483.  
538 <https://doi.org/10.1017/S0025315408000945>
- 539 De Carolis, B., Derosis, F., Grasso, F., Rossiello, A., Berry, D. C., & Gillie, T. (1996).  
540 Generating recipient-centered explanations about drug prescription. *Artif. Intell.*  
541 *Med.* 8: 123–145. [https://doi.org/10.1016/0933-3657\(95\)00029-1](https://doi.org/10.1016/0933-3657(95)00029-1)
- 542 De Donno, A., Idolo, A., Bagordo, F., Grassi, T., Leomanni, A., Serio, F., Guido, M.,  
543 Canitano, M., Zampardi, S., Boero, F., & Piraino, S. (2014). Impact of stinging  
544 jellyfish proliferations along south Italian coasts: human health hazards,  
545 treatment and social costs. *International journal of environmental research and*  
546 *public health* 11(3): 2488-2503. <https://doi.org/10.3390/ijerph110302488>
- 547 Delannoy, C. M. J., Houghton, J. D. R., Fleming, N. E. C., & Ferguson, H. W. (2011).  
548 Mauve Stingers (*Pelagia noctiluca*) as carriers of the bacterial fish pathogen  
549 *Tenacibaculum maritimum*. *Aquaculture* 311(1-4): 255-257.  
550 <https://doi.org/10.1016/j.aquaculture.2010.11.033>
- 551 Dong, Z., Liu, D., & Keesing, J. K. (2010). Jellyfish blooms in China: dominant  
552 species, causes and consequences. *Marine pollution bulletin* 60(7): 954-963.  
553 <https://doi.org/10.1016/j.marpolbul.2010.04.022>
- 554 Doyle T. K., DeHaas, H., Cotton, D., Dorschel, B., Cummins, V., Houghton, J. D. R.,  
555 Davenport, J., & Hays, G. C. (2008). Widespread occurrence of the  
556 jellyfish *Pelagia noctiluca* in Irish coastal and shelf waters. *J. Plankton. Res.*  
557 30(8): 963–968. <https://doi.org/10.1093/plankt/fbn052>

- 558 Duarte, C. M., Pitt, K. A., & Lucas, C. H. (2014). Introduction: Understanding jellyfish  
559 blooms. In K. A. Pitt & C. H. Lucas (Eds.), *Jellyfish blooms* (pp. 1-5). Springer  
560 Science+Business Media.
- 561 Enríquez, J. G, Cid, V., Muntaner, N., Aroba, J., Navarro, J., Domínguez, F. J.,  
562 Escalona, M. J., & Ramos, I. (2018). Behavior patterns in hormonal treatments  
563 using fuzzy logic-based models. *Soft Computing* 22: 79-90.  
564 <https://doi.org/10.1007/s00500-017-2614-7>
- 565 Fernández, A., Herrera, F., Cordón, O., del Jesús, M. J., & Marellonu, F. (2019).  
566 Evolutionary fuzzy systems for explainable artificial intelligence: Why, when,  
567 what for, and where to?, *IEEE Computational Intelligence Magazine* 14 (1): 69-  
568 81. <https://doi.org/10.1109/mci.2018.2881645>
- 569 Fernández-Camacho, R., Brito Cabeza, I., Aroba, J., Gómez-Bravo, F., Rodríguez, S., &  
570 de la Rosa, J. (2015). Assessment of ultrafine particles and noise measurements  
571 using fuzzy logic and data mining techniques. *Journal of Science of the Total*  
572 *Environment* 512-513: 103-113. <https://doi.org/10.1016/j.scitotenv.2015.01.036>
- 573 Fossette, S., Gleiss, A. C., Chalumeau, J., Bastian, T., Armstrong, C. D., Vandenabeele,  
574 S., Karpytchev, M., & Hays, G. C. (2015). Current-Oriented Swimming by  
575 Jellyfish and Its Role in Bloom Maintenance. *Current Biology* 25(3): 342-347.  
576 <https://doi.org/10.1016/j.cub.2014.11.050>
- 577 Fukuyama, Y., & Sugeno, M. (1989). A new method of choosing the number of clusters  
578 for fuzzy means method. *Proceedings of the 5th Fuzzy Systems Symposium* (pp.  
579 247-250).
- 580 Gangopadhyay, A. (2001). Conceptual modeling from natural language functional  
581 specifications. *Artif. Intell. Eng.* 15(2): 207–218. [https://doi.org/10.1016/S0954-](https://doi.org/10.1016/S0954-1810(01)00017-6)  
582 [1810\(01\)00017-6](https://doi.org/10.1016/S0954-1810(01)00017-6)
- 583 García, E., & Ocaña, C. (1982). La organización espacial de la Costa Mediterránea  
584 Andaluza. *Baetica* 5: 15-57.
- 585 García-Lafuente, J., Sánchez-Garrido, J. C., Garcia, A., Hidalgo, M., Sammartino, S., &  
586 Laíz, R. (2021). Biophysical Processes Determining the Connectivity of the

- 587 Alboran Sea Fish Populations. In J.C. Báez, J.T. Vázquez, J. T., J.A. Camiñas, &  
588 M.M. Idrissi (Eds.), *Alboran Sea - Ecosystem and Marine Resources* (pp. 459-  
589 487). Springer Nature.
- 590 García-Soto C, Seys, J. J. C., Zielinski, O., Busch, J. A., Luna, S. I., Baez, J. C.,  
591 Domegan, C., Dubsky, K., Kotynska-Zielinska, I., Loubat, P., Malfatti, F.,  
592 Mannaerts, G., McHugh, P., Monestiez, P., van der Meeren, G.I., & Gorsky, G.  
593 (2021). Marine Citizen Science: Current State in Europe and New Technological  
594 Developments. *Front. Mar. Sci.* 8: 621472.  
595 <https://doi.org/10.3389/fmars.2021.621472>
- 596 Ghermandi, A., Galil, B., Gowdy, J., & Nunes, P. A. L. D. (2015). Jellyfish outbreak  
597 impacts on recreation in the Mediterranean Sea: welfare estimates from a  
598 socioeconomic pilot survey in Israel. *Ecosyst. Serv.* 11: 140–147.  
599 <https://doi.org/10.1016/j.ecoser.2014.12.004>
- 600 Graham, W. M., Martin, D. L., Felder, D. L., Asper, V. L., & Perry, H. M. (2003).  
601 Ecological and economic implications of a tropical jellyfish invader in the Gulf  
602 of Mexico. *Biological Invasions* 5: 53–69.  
603 <https://doi.org/10.1023/A:1024046707234>
- 604 Graham, W. M., Pagès, F., & Hamner, W. M. (2001). A physical context for gelatinous  
605 zooplankton aggregations: a review. *Hydrobiologia* 451: 199–212.
- 606 Gutiérrez-Estrada, J. C., De Pedro Sanz, E., López-Luque, R., & Pulido-Calvo, I.  
607 (2005). SEDPA, an expert system for disease diagnosis in eel rearing systems.  
608 *Aquacult. Eng.* 33: 110–125. <https://doi.org/10.1016/j.aquaeng.2004.12.003>
- 609 Gutiérrez-Estrada, J. C., Pulido-Calvo, I., de la Rosa, I., & Marchini, B. (2012).  
610 Modelling inflow rates for the water exchange management in semi-intensive  
611 aquaculture ponds. *Aquacultural Engineering* 48: 19-30.  
612 <https://doi.org/10.1016/j.aquaeng.2011.12.009>
- 613 Gutiérrez-Estrada, J. C., Pulido-Calvo, I., Peregrín, A., García-Gálvez, A., Báez, J. C.,  
614 Bellido, J. J., Souviron-Priego, L., Sánchez-Laulhé, J. M., & López, J. A.  
615 (2021). Integrating local environmental data and information from non-driven

- 616 citizen science to estimate jellyfish abundance in Costa del Sol (southern  
617 Spain). *Estuarine, Coastal and Shelf Science* 249: 107112.  
618 <https://doi.org/10.1016/j.ecss.2020.107112>
- 619 Gutiérrez-Estrada, J. C., Pulido-Calvo, I., & Bilton, D. T. (2013). Consistency of fuzzy  
620 rules in an ecological context. *Ecological Modelling* 251: 187-198.  
621 <https://doi.org/10.1016/j.ecolmodel.2012.12.013>
- 622 Halsband, C., Majaneva, S., Hosia, A., Emaus, P. A., Gaardsted, F., Zhou, Q., Nøst, O.  
623 A., & Renaud, P. E. (2018). Jellyfish summer distribution, diversity and impact  
624 on fish farms in a Nordic fjord. *Marine Ecology Progress Series* 591: 267-279.  
625 <https://doi.org/10.3354/meps12274>
- 626 Hamner, W. M., & Dawson, M. N. (2009). A review and synthesis on the systematics  
627 and evolution of jellyfish blooms: advantageous aggregations and adaptive  
628 assemblages. *Hydrobiologia* 616: 161-191. [https://doi.org/10.1007/s10750-008-](https://doi.org/10.1007/s10750-008-9620-9)  
629 [9620-9](https://doi.org/10.1007/s10750-008-9620-9)
- 630 Hay, S. (2006). Marine ecology: gelatinous bells may ring change in marine  
631 ecosystems. *Current Biology* 16: R679–R682.  
632 <https://doi.org/10.1016/j.cub.2006.08.010>
- 633 Heinle, D. R. (1965). A screen for excluding jellyfish and ctenophores from Clarke-  
634 Bumpus plankton samples. *Chesapeake Science* 6: 231–232
- 635 Höppner, F., & Klawonn, F. (2003). A contribution to convergence theory of fuzzy c-  
636 means and derivatives. *IEEE Trans. Fuzzy Syst.* 11: 682-694.  
637 <https://doi.org/10.1109/TFUZZ.2003.817858>.
- 638 Knowler, D. (2005). Reassessing the costs of biological invasion: *Mnemiopsis leidyi* in  
639 the Black Sea. *Ecological Economics* 52(2): 187-199.  
640 <https://doi.org/10.1016/j.ecolecon.2004.06.013>
- 641 Komninos, A. (2018). Pro-social Behaviour in Crowdsourcing Systems: Experiences  
642 from a field deployment for beach monitoring. *International Journal of Human-*  
643 *Computer Studies* 124: 93-115. <https://doi.org/10.1016/j.ijhcs.2018.12.001>.

- 644 Lee, T., Quinn, M., & Duke, D. (2006). Citizen, Science, High-ways, and Wildlife:  
645 Using a Web-based GIS to Engage Citizens in Collecting Wildlife Information.  
646 *Ecology and Society* 11(1): 11.
- 647 Luis, A. T., Grande, J. A., Dávila, J. M., Aroba, J., Duraes, N., Almeida, S. F., de la  
648 Torre, M. L., Sarmiento, A. M., Fortes, J. C., Ferreira da Silva, E., &  
649 Santisteban, M. (2018). Application of fuzzy logic tools for the biogeochemical  
650 characterisation of (un)contaminated waters from Aljustrel mining area (South  
651 Portugal). *Chemosphere* 211: 736-744.  
652 <https://doi.org/10.1016/j.chemosphere.2018.07.194>
- 653 Macías, D., García-Gorriz, E., & Stips, A. (2016). The seasonal cycle of the Atlantic Jet  
654 dynamics in the Alboran Sea: direct atmospheric forcing versus Mediterranean  
655 thermohaline circulation. *Ocean Dynamics* 66(2): 137-151.  
656 <https://doi.org/10.1007/s10236-015-0914-y>
- 657 Malej, A. (1989). Behaviour and trophic ecology of the jellyfish *Pelagia noctiluca*  
658 (Forsskål, 1775). *Journal of Experimental Marine Biology and Ecology* 126(3):  
659 259-270. [https://doi.org/10.1016/0022-0981\(89\)90191-3](https://doi.org/10.1016/0022-0981(89)90191-3)
- 660 Martín del Brío, B., & A. Sanz-Molina. 2006. *Redes neuronales y sistemas borrosos.*  
661 Ra-Ma.
- 662 Mercado, J. M., Cortés, D., Ramírez, T., & Gómez, F. (2012). Decadal weakening of  
663 the wind-induced upwelling reduces the impact of nutrient pollution in the Bay  
664 of Málaga (western Mediterranean Sea). *Hydrobiologia* 680: 91–107.  
665 <https://doi.org/10.1007/s10750-011-0906-y>
- 666 Nagata, R. M., Haddad, M. A., & Nogueira, M. (2009). The nuisance of medusae to  
667 shrimp trawls in central part of southern Brazilian Bight, from the perspective of  
668 artisanal fishermen. *Pan-American Journal of Aquatic Sciences* 4: 312–325.
- 669 Nastav, B., Malej, M., Malej Jr, A., & Malej, A. (2013). Is it possible to determine the  
670 economic impact of jellyfish outbreaks on fisheries? A case study–  
671 Slovenia. *Mediterranean Marine Science* 14(1): 214-223.  
672 <https://doi.org/10.12681/mms.382>

- 673 Peliz, A., Boutov, D., & Teles-Machado, A. (2013). The Alboran Sea mesoscale in a  
674 long term high resolution simulation: Statistical analysis. *Ocean Modelling* 72: 32-52.  
675 <https://doi.org/10.1016/j.ocemod.2013.07.002>
- 676 Pugh, P. R. (1989). Gelatinous zooplankton – the forgotten fauna. *Progress in*  
677 *Underwater Science* 14: 67–78
- 678 Purcell, J. E. (2005). Climate effects on formation of jellyfish and ctenophore blooms: a  
679 review. *Journal of the Marine Biological Association of the United*  
680 *kingdom* 85(3): 461-476.
- 681 Purcell, J. E. (2012). Jellyfish and ctenophore blooms coincide with human  
682 proliferations and environmental perturbations. *Annual review of marine*  
683 *science* 4: 209-235.
- 684 Ramírez, T., Muñoz, M., Reul, A., García-Martínez, M. C., Moya, F., Vargas-Yáñez,  
685 M., & Bautista, B. (2021). The Biogeochemical Context of Marine Planktonic  
686 Ecosystems. In J. C. Báez, J. T. Vázquez, J. A. Camiñas, M. M. Idrissi (Eds.),  
687 *Alboran Sea - Ecosystem and Marine Resources* (pp. 207-246). Springer Nature.
- 688 Rodríguez-García, C., Sanz-Fernández, V., Muñoz-Lechuga, R., Gutiérrez-Martínez,  
689 M., & Cabrera-Castro, R. (2021). First records and presence over time of the  
690 cubozoan *Carybdea marsupialis* (Linnaeus, 1758) on the southwestern Spanish  
691 coast (Northeast Atlantic). *Regional Studies in Marine Science* 44: 101712.  
692 <https://doi.org/10.1016/j.rsma.2021.101712>
- 693 Romero-Padilla, Y., Navarro-Jurado, E., & Romero-Martínez, J-M. (2020). Tourist  
694 destinations and creative capital: The case of costa del sol in southern  
695 Spain. *Revista de geografía Norte Grande* 77: 339-  
696 365. <https://dx.doi.org/10.4067/S0718-34022020000300339>
- 697 Sugeno, M., & Yasukawa, A. (1993). A fuzzy-logic based approach to qualitative  
698 modeling. *IEEE Trans. Fuzzy Syst.* 1(1): 7-31.
- 699 Tang, T. -C., & Chi, L. -C. (2005). Predicting multilateral trade credit risks:  
700 comparisons of Logit and Fuzzy logic-based models using ROC curve analysis.

- 701 *Expert Systems with Applications* 28(3): 547–  
702 556. <https://doi.org/10.1016/j.eswa.2004.12.016>
- 703 Templado, J., Luque, Á. A., Moreno, D., de Figueroa, J. M. T., Tocino, L. S., Aguilar,  
704 R., & de la Torriente, A. (2021). Invertebrates: The Realm of Diversity. In J. C.  
705 Báez, J. T. Vázquez, J. A. Camiñas, M. M. Idrissi (Eds.), *Alboran Sea -*  
706 *Ecosystem and Marine Resources* (pp. 359-430). Springer Nature.
- 707 Vargas-Yáñez, M., García-Martínez, M. C., Moya, F., Balbín, R., & López-Jurado, J. L.  
708 (2021). The Oceanographic and Climatic Context. In J. C. Báez, J. T. Vázquez,  
709 J. A. Camiñas, M. M. Idrissi (Eds.), *Alboran Sea - Ecosystem and Marine*  
710 *Resources* (pp. 85-109). Springer Nature.
- 711 Woods, W. A. (1970). Transition networks grammars for natural language analysis.  
712 *Commun. Assoc. Comput. Mach.* 13(10): 591–610.
- 713 Woods, W. A. (1973). Progress in natural language understanding: an application to  
714 Lunar geology. In *Proceedings of the June 4-8, 1973, National computer*  
715 *conference and exposition, AFIPS '73* (pp. 441–450). ACM.
- 716 Yi, Y., Sun, J., Zhang, S., & Yang, Z. (2016). Assessment of Chinese sturgeon habitat  
717 suitability in the Yangtze River (CHINA): Comparison of generalized additive  
718 model, data-driven fuzzy logic-based model, and preference curve model.  
719 *Journal of Hydrology* 536: 447-456.  
720 <https://doi.org/10.1016/j.jhydrol.2016.03.005>
- 721 Zamora, W., Calafate, C. T., Cano, J. -C., & Manzoni, P. (2016). A survey on  
722 smartphone-based crowdsensing solutions. *Mob. Inf. Syst.* 2016: 26  
723 <https://doi.org/10.1155/2016/9681842>
- 724 Zavodnik, D. (1987). Spatial aggregations of the swarming jellyfish *Pelagia*  
725 *noctiluca* (Scyphozoa). *Mar Biol* 94: 265–269.