

1 **Fishers' local knowledge strengthens seagrass restoration planning**

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9 **Abstract**

10 Seagrass restoration is increasingly guided by habitat suitability models, yet restoration
11 outcomes depend on more than biophysical suitability alone. In coastal social-ecological
12 systems, fishers and anglers hold fine-scale, time-integrated knowledge of habitat
13 condition, human use, and local constraints that are rarely incorporated at the outset of
14 restoration planning. Here, we tested whether fisher and angler knowledge could generate
15 a spatially explicit basis for seagrass restoration site selection along the southern Welsh
16 coastline, UK. Using participatory mapping and a Wisdom of Crowds approach, 33 coastal
17 resource users identified locations perceived as suitable for seagrass restoration and areas
18 where restoration should be avoided. Weighting responses by regional participation and
19 converting them into kernel density surfaces, we tested whether participatory-derived
20 suitability predicted independently observed seagrass occurrence and assessed
21 stakeholder perceptions of seagrass benefits. Participatory mapping revealed coherent
22 hotspots of perceived restoration opportunity in sheltered bays and estuarine environments,
23 whereas avoidance areas clustered around ports, sediment-influenced estuaries, and high-
24 use tourism locations, indicating that participants integrated both ecological opportunity
25 and human-use constraints. Net suitability was a strong predictor of seagrass occurrence,
26 demonstrating close correspondence between aggregated local knowledge and observed
27 seagrass distribution. Perceptions of seagrass restoration were broadly positive, with strong
28 agreement that seagrass benefits marine life and the wider environment. Our findings show
29 that fisher and angler knowledge can generate spatially coherent and ecologically
30 meaningful information for restoration planning. Rather than acting as a simple substitute
31 for habitat suitability modelling, participatory mapping functions as a social-ecological
32 diagnostic layer, identifying areas of opportunity, constraint, and potential conflict.
33 Integrating local knowledge early in restoration planning can improve site selection, guide
34 choices between protection, passive recovery and active restoration, and strengthen
35 legitimacy, stewardship, and long-term support for seagrass recovery.

36 Keywords: Participatory mapping; Stakeholder engagement; Social-ecological systems;
37 Knowledge co-production; Local ecological knowledge; Seagrass restoration; Restoration
38 planning; Habitat suitability

39

40

41 **Introduction**

42 Environmental decision-making often proceeds under uncertainty, shaped by data scarcity,
43 coarse spatial resolution, and incomplete understanding of ecological and social dynamics
44 (Gregory et al., 2012; Milner-Gulland and Shea, 2017). Indigenous and Local Knowledge (ILK)
45 is increasingly recognised as a critical evidence base for terrestrial and marine management,
46 particularly where empirical data are sparse, costly, or spatiotemporally limited (Brook and
47 McLachlan, 2008; Stern and Humphries, 2022; Jones et al., 2024). Across ecological
48 disciplines, ILK has been used to infer species distributions, reconstruct historical baselines,
49 and inform habitat suitability and species distribution models (Smith et al., 2007; Lopes et
50 al., 2019; Early-Capistran et al., 2020). Participatory mapping and interview-based
51 approaches with resource users have also been successfully integrated with spatial
52 modelling frameworks, often improving predictions of habitat extent and ecological change
53 (Stern and Humphries, 2022; Jones et al., 2024).

54 The “Wisdom of Crowds” provides a complementary theoretical foundation for this
55 approach. Under the right conditions, aggregated estimates from diverse and independent
56 individuals can match or exceed expert judgement (Surowiecki, 2004; Kattan et al., 2016;
57 Wu et al., 2020; Turiel et al., 2021; Jones et al., 2025b). Applied to ecological systems, this
58 suggests that experience-based knowledge, built through repeated interaction with natural
59 environments, can generate robust insights into ecological structure, function, and change.
60 In practice, structured elicitation and spatial aggregation of ILK have revealed patterns that
61 are either consistent with, or entirely absent from, ecologically modelled outputs (Whitmore,
62 2016; Gray et al., 2020; Jones et al., 2025b).

63 Despite this potential, ILK remains unevenly integrated into formal spatial modelling and
64 conservation planning. In marine systems, Habitat Suitability Models (HSMs) and Species
65 Distribution Models (SDMs) are commonly driven by environmental covariates derived from
66 remote sensing or interpolated datasets (Monk et al., 2011; Souza Oliveira et al., 2025).
67 These approaches are powerful, but their reliability depends entirely on the resolution,
68 availability, and assumptions of their input layers, with many models utilising open-source
69 data. As a result, model reliability, robustness, and performance are constrained by the
70 resolution, quality and ecological relevance of available predictors across space and time
71 (Barry and Elith, 2006; Cushway et al., 2026). Furthermore, fine-scale heterogeneity,
72 particularly in dynamic coastal environments, may be systematically underrepresented.

73 These limitations matter because HSMs are increasingly central to restoration planning.
74 They are widely used to identify priority areas, guide site selection, and optimise intervention
75 design (Questad et al., 2014; Yuen et al., 2023; Shyvers et al., 2024). Yet restoration is not
76 simply an ecological intervention. In social–ecological systems (SES)(Ostrom, 2009),

77 restoration outcomes emerge through interactions among ecological conditions,
78 governance, knowledge systems, resource users, and local values (Tedesco et al., 2023).
79 Knowledge acquisition, production, and implementation are therefore core components of
80 restoration planning, rather than external inputs (Unsworth et al., 2026b), indeed as it
81 should be (Di Sacco et al., 2021; Quigley et al., 2022; Unsworth et al., 2024). Seagrass
82 meadows provide a compelling SES through which to explore these components.

83 Globally, seagrass ecosystems underpin coastal productivity and resilience (Unsworth et
84 al., 2022), sequestering carbon, stabilising sediments, attenuating wave energy, improving
85 water quality, and providing vital habitat for a wide range of species. Critically, they underpin
86 fisheries by providing nursery and juvenile habitat for numerous species of finfish and
87 shellfish (Unsworth et al., 2018), linking ecosystem function directly to human livelihoods,
88 particularly in subsistence fishery reliant communities (Jones et al., 2021; Jones et al.,
89 2022b). In the UK, seagrass habitats support commercially and recreationally important fish
90 species, such as Atlantic Cod (*Gadus morhua*)(Bertelli and Unsworth, 2014), and in other
91 temperate systems, seagrass can nearly double cockle (*Cerastoderma edule*) feeding rates
92 (Brun et al., 2009).

93 Despite their ecological and socio-economic importance, seagrass meadows are in global
94 decline (Dunic et al., 2021), threatened by pollution, coastal development and climate
95 change (Jones et al., 2025a). Restoration efforts have expanded in response (van Katwijk et
96 al., 2016), but current approaches remain largely model-driven. Restoration site selection
97 typically draws on HSM outputs, ground-truthing and experimentation (Stankovic et al.,
98 2019; Bertelli et al., 2022; Grigg et al., 2025), with stakeholder engagement often introduced
99 later to manage conflict, support implementation, or build local acceptance. This
100 sequencing creates two limitations. First, it risks over-reliance on model assumptions and
101 available datasets, which could lead to poor restoration outcomes when unsuitable sites
102 are selected (Coals et al., 2025). Second, it misses opportunities to incorporate locally held
103 ecological knowledge and social context early in the process (Smith et al., 2022). The result
104 is not only a potential gap in ecological understanding, but also a risk of weakening
105 community ownership, stewardship, and broader support (Dennis, 2025).

106 Fishers and anglers hold detailed, place-based knowledge accumulated through sustained
107 engagement with coastal environments, often over decades and across generations (Jones
108 et al., 2024; Mason et al., 2025). A potential benefit of this time-integrated knowledge is that
109 it overcomes many of the problems associated with single time point datasets inherent in
110 the creation of HSMs. Such knowledge is well positioned to support seagrass restoration
111 planning because it can capture local ecological conditions, historical baselines, temporal
112 change, and location-specific pressures that may be difficult to detect in conventional

113 datasets (Berkström et al., 2019; Jones et al., 2022a). When systematically aggregated
114 (Jones et al., 2025b), fisher and angler knowledge also reflects the principles of the “Wisdom
115 of Crowds”, offering a potentially robust and scalable source of spatial evidence for
116 restoration planning.

117 However, spatial knowledge alone is insufficient for understanding restoration potential
118 within an SES. Restoration also depends on how stakeholders perceive ecological benefits,
119 resource benefits, and personal relevance. Such perceptions shape whether restoration is
120 viewed as desirable, legitimate, and worth supporting, and they influence how resource
121 users interpret what restoration is for (Matzek and Wilson, 2021; Löfqvist et al., 2023).
122 Measuring these perceptions alongside spatial knowledge can therefore reveal whether
123 mapped restoration opportunities are underpinned by broader social support, uncertainty,
124 or contested values.

125 Here, we test whether aggregated fisher and angler knowledge can identify suitable
126 seagrass restoration sites along the southern Welsh coastline, UK (Unsworth et al., 2026a).
127 Rather than treating ILK as supplementary, we position it as an alternative, or
128 complementary, approach to conventional HSMs. Specifically, we assess how
129 participatory-derived site suitability maps compare with independently observed seagrass
130 data, and how stakeholder perceptions of seagrass and restoration benefits align with these
131 spatial patterns. In doing so, we evaluate whether experiential knowledge can meaningfully
132 inform restoration planning within this SES. While inherently place-based, this approach
133 provides a framework transferable to other local, regional and global systems where
134 ecological conditions and knowledge contexts are comparable.

135

136 **Methods**

137 *Study context*

138 The study was conducted along the southern coastline of Wales, UK. This coastline,
139 hereafter referred to as south Wales for simplicity, spans from Pembrokeshire in the west,
140 to Monmouthshire in the east, and encompasses a continuum of estuaries, embayments,
141 and shallow coastal systems that have undergone systematic anthropogenic induced
142 change over a period of 300 years (Unsworth et al., 2026a). The study region also spans three
143 ICES regional seas; Bristol Channel, Celtic Sea and Irish Sea (Moore et al., 2023). While
144 regional definitions vary in policy and common usage, the south Wales framing reflects a
145 semi-continuous social-ecological system linked by coastal processes, fisheries activity,
146 and shared cultural identity.

147 The region supports a long-standing and diverse suite of coastal fisheries situated across
148 three ICES regional seas; Bristol Channel, Celtic Sea and Irish Sea (Moore et al., 2023). The
149 region remains characterised by active, highly engaged fishing and angling communities
150 (Monkman et al., 2018; Silva and Ellis, 2019). It supports a mosaic of commercial,
151 recreational, and traditional fisheries centred on estuarine and nearshore systems, most
152 notably the Burry Inlet cockle fishery, which has operated for over a century and remains
153 one of the most significant shellfisheries in Wales (Norris et al., 1998). These fisheries are
154 embedded within local economies and livelihoods, with hand-gathering methods and
155 management systems reflecting strong continuity across generations. More broadly, the
156 south Wales coast sustains a range of small-scale and recreational fisheries that interact
157 with dynamic coastal habitats, including estuaries and shallow embayments where
158 seagrass meadows occur (Monkman et al., 2018). While some fisheries have declined or
159 shifted in form over time, long-term monitoring and historical analyses indicate persistent
160 ecological and socio-economic linkages between shellfisheries, environmental change,
161 and coastal management in the region (Callaway, 2022).

162 The region also includes prominent seagrass restoration initiatives, most notably Dale in
163 Pembrokeshire, widely recognised as the first large-scale seagrass restoration site in the UK
164 (Unsworth et al., 2026b), as well as extensive historic seagrass meadows across the Milford
165 Haven area and the Burry Inlet (Jones and Unsworth, 2016; Bertelli et al., 2018). The Welsh
166 Government has also made a commitment to seagrass protection and restoration, investing
167 in the National Seagrass Action Plan. As such, the region provides a relevant and timely
168 context in which to evaluate alternative approaches to restoration planning, alongside
169 ongoing national efforts to scale seagrass recovery.

170 *Stakeholder engagement*

171 An initial pilot study was conducted through an in-person, map-based workshop at two
172 Seagrass Nursery Open Days in Laugharne, Carmarthenshire. Forty-two participants were
173 invited to map known seagrass presence, identify areas they believed suitable for seagrass
174 growth, and highlight locations where restoration should be avoided by identifying
175 anthropogenic pressures such as pollution or conflicting activities (*i.e.*, bait digging).
176 Participants were invited from a mixed stakeholder group, based on their interest in local
177 seagrass restoration, and included members of local community science groups, regulatory
178 bodies, NGOs and local businesses. This exercise was supported by a facilitated discussion
179 on the ecological conditions required for seagrass establishment. Workshop outputs were
180 subsequently digitised for analysis.

181 Several challenges, however, arose during the workshop. Participants varied in their
182 familiarity with the coastline presented, and some lacked regular direct engagement with

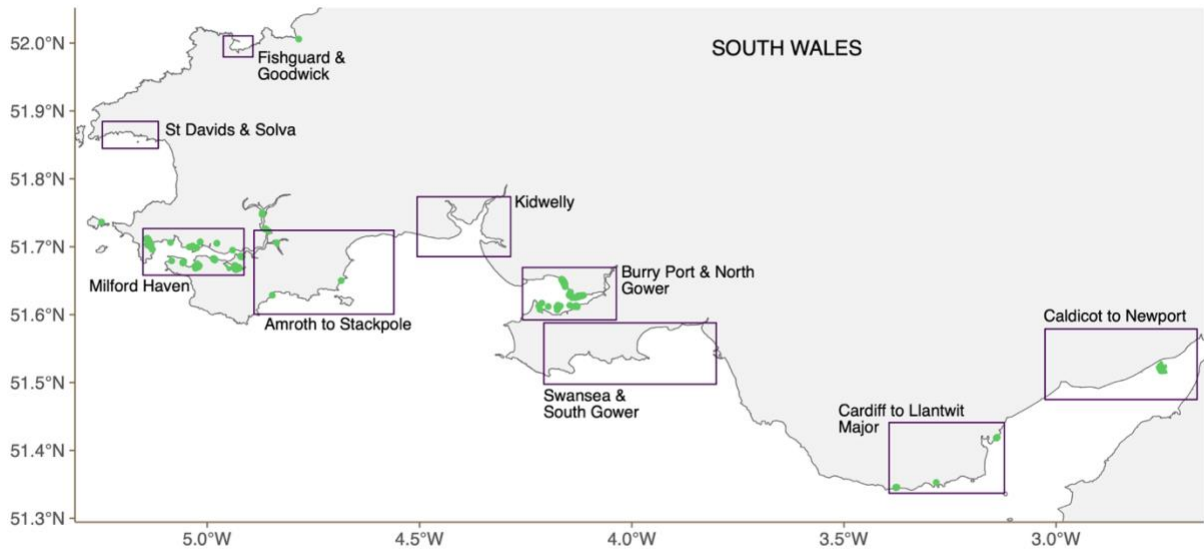
183 the areas in question. The in-person format also constrained stakeholder participation,
184 excluding individuals unable to attend due to time or travel limitations, and was biased
185 towards individuals with prior knowledge of seagrass restoration and/or whom had a level
186 of interest sufficient to motivate attendance at a seagrass-related event. Additionally, the
187 group setting may have introduced further bias, as participants could see and be potentially
188 influenced by others' responses (O.Nyumba et al., 2018), limiting the independence
189 necessary for robust knowledge aggregation.

190 To address these limitations, a structured online survey was developed, specifically
191 targeting fishers and anglers, given their sustained interaction with marine and coastal
192 environments and detailed local knowledge. The online format enabled broader geographic
193 reach and greater participation flexibility, with surveys designed to be completed in a
194 reasonable timeframe *i.e.* within 10–15 minutes. Rather than assuming seagrass knowledge
195 *a priori*, all participants were provided with a clear description of seagrass morphology and
196 ecology associated with *Zostera marina*, *Nanozostera noltii* and *Ruppia maritima* before
197 taking the survey.

198 To capture social-ecological variation, the south Wales coastline was divided into sub-
199 regions reflecting differences in seagrass species composition, geographic clustering and
200 patterns of resource use (Figure 1). For example, some areas are characterised by intertidal
201 *N. Noltii* and *R. maritima*, often associated with cockle beds and saltmarsh systems, while
202 others are dominated by subtidal *Z.marina*, supporting fisheries targeting species such as
203 crab and lobster. In the online survey, interactive maps for each sub-region allowed
204 respondents to spatially identify up to five areas of perceived suitability and five areas of
205 unsuitability, alongside questions to share their local insights and perceptions of seagrass
206 restoration more broadly. Importantly, participants were asked which fishing region they
207 visited most, and only that map was provided to participants.

208 The online survey was circulated during the winters of 2024/5 and 2025/6 across the south
209 Wales area, with the intention of targeting commercial fishers, shellfish harvesters,
210 aquaculturists, recreational anglers and spearfishers. To maximise diversity of responses,
211 consistent with the logic of a Wisdom of Crowds approach, the survey was distributed
212 through a multi-pronged recruitment strategy combining purposive, opportunistic, and
213 snowball sampling. Recruitment routes included: QR-coded leaflets placed in bait and
214 tackle shops, email circulation through fishing networks, paid social media advertisements,
215 posts in fishing-related Facebook groups, word-of-mouth sharing, direct engagement with
216 fishers and anglers, dissemination through local nature partnerships, and coverage in local
217 press. Responses were then collated and analysed to investigate patterns in site selection,

218 perceptions of the community, and to compare the sites selected through the surveys with
219 those selected through habitat suitability modelling and ground-truthing.



220
221 **Figure 1.** South Wales study region showing the nine coastal regions used in the participatory mapping
222 survey. Purple boxes indicate the survey map areas presented to participants, while green points and
223 polygons show existing seagrass records (Green et al., 2021; Rice et al., 2022; Jones et al., 2025a),
224 including mapped meadows, point observations, and fragment sightings.

225
226 Responses from 33 of 42 participants were retained for analysis, based on either full survey
227 completion or provision of sufficient spatial data. These responses generated a
228 georeferenced dataset of participant-selected locations, classified as either suitable for
229 seagrass restoration or areas where restoration should be avoided. These data were then
230 used to produce spatial heat maps of perceived restoration opportunity and constraint
231 across the study regions.

232 *Data analysis*

233 Survey data were imported and processed in R (R Core Team, 2026). For the mapping data,
234 each respondent was assigned a unique *stakeholder_id*, and their reported primary fishing
235 region was retained for subsequent weighting. Given spatial responses were originally
236 stored as paired longitude–latitude coordinates across multiple ranked choices (e.g.
237 “Suitable.Choice.1.x”, “Suitable.Choice.1.y”), these were reshaped to create two distinct
238 datasets representing locations identified as suitable and those identified as unsuitable

239 (avoid), which were subsequently combined into a single dataset with a categorical variable
240 distinguishing response type. Records with missing coordinates were removed.

241 Next, to reduce spatial bias arising from uneven sampling effort across regions, responses
242 were weighted inversely by the number of respondents per region. Specifically, each point
243 was assigned a weight:

$$244 \quad \text{weight} = \frac{1}{n_{region}}$$

245
246 This ensured that geographic areas with higher participation did not disproportionately
247 influence the analysis.

248 Resulting point data were converted to a spatial object using the *sf* package for R (Pebesma,
249 2018), and projected into the British National Grid coordinate reference system
250 (EPSG:27700) to allow for spatial analyses requiring metric distances, such as Kernel
251 Density Estimation (KDE) (Baddeley et al., 2016). Projected coordinates were then converted
252 into a planar point pattern object using the *spatstat* package, with point weights retained as
253 marks (Baddeley et al., 2016).

254 Given that spatial responses were collected using a mobile-compatible map interface using
255 Qualtrics, point locations were treated as spatially uncertain observations rather than
256 precise coordinates. The extent of each map presented to respondents varied between
257 regions (5–27 km), and responses were typically recorded on mobile devices (screen widths
258 ~ 6.5–7.5 cm). To account for this, positional uncertainty was estimated by converting map
259 extent to a spatial resolution (km per cm), and applying a conservative tap error of 0.5 cm,
260 representing the approximate precision of a finger-based input. This produced region-
261 specific uncertainty distances for each point. Rather than buffering points geometrically,
262 these uncertainty estimates were incorporated directly into the KDE as smoothing
263 bandwidths, allowing each observation to contribute a spatially diffuse signal consistent
264 with its expected positional precision.

265 Weighted kernel density estimation was applied using Gaussian kernels, incorporating both
266 respondent weights and spatial uncertainty (Diggle, 1985). Bandwidths were defined based
267 on the median estimated positional uncertainty for each response type (suitable and avoid),
268 ensuring that smoothing reflected the scale of input error rather than relying solely on data-
269 driven optimisation. Edge correction was applied to reduce boundary effects.

270 Finally, each density surface was normalised independently by dividing by its maximum
271 value, resulting in continuous surfaces scaled from 0 to 1. This standardisation was used to
272 allow direct comparison between suitability and avoidance intensities.

273 We then derived a net suitability surface as:

274

275
$$\text{Net suitability} = \text{Suitability density} - \text{Avoidance density}$$

276

277 This produces a continuous index ranging from -1 to 1 where positive values indicate areas
278 perceived as suitable, negative values indicate areas perceived as unsuitable, and values
279 near zero indicate neutrality, or disagreement among respondents. We first normalised this
280 to a scale of 0-1, then to improve interpretability, near-zero values (0.475 to 0.525) were set
281 to missing, effectively masking areas of low consensus.

282 To evaluate the relationship between participatory suitability mapping and independently
283 recorded seagrass occurrence, external presence data were obtained from
284 SeagrassSpotter records (Jones et al., 2025a) and other published seagrass point data
285 (Green et al., 2021). We extracted net suitability values at each seagrass observation
286 location using the *tterra* package. Next, to provide a baseline for comparison, we generated
287 a set of randomly distributed background points across the study extent using spatial
288 random sampling, and net suitability values were extracted for these locations using the
289 same procedure.

290 The relationship between suitability and seagrass occurrence was first assessed visually
291 using density plots. To formally test predictive performance, binomial logistic regression
292 models were fitted with presence-background status as the response variable and
293 suitability as the predictor. Model outputs were used to generate continuous prediction
294 functions, and model performance was evaluated using the area under the receiver
295 operating characteristic curve (AUC).

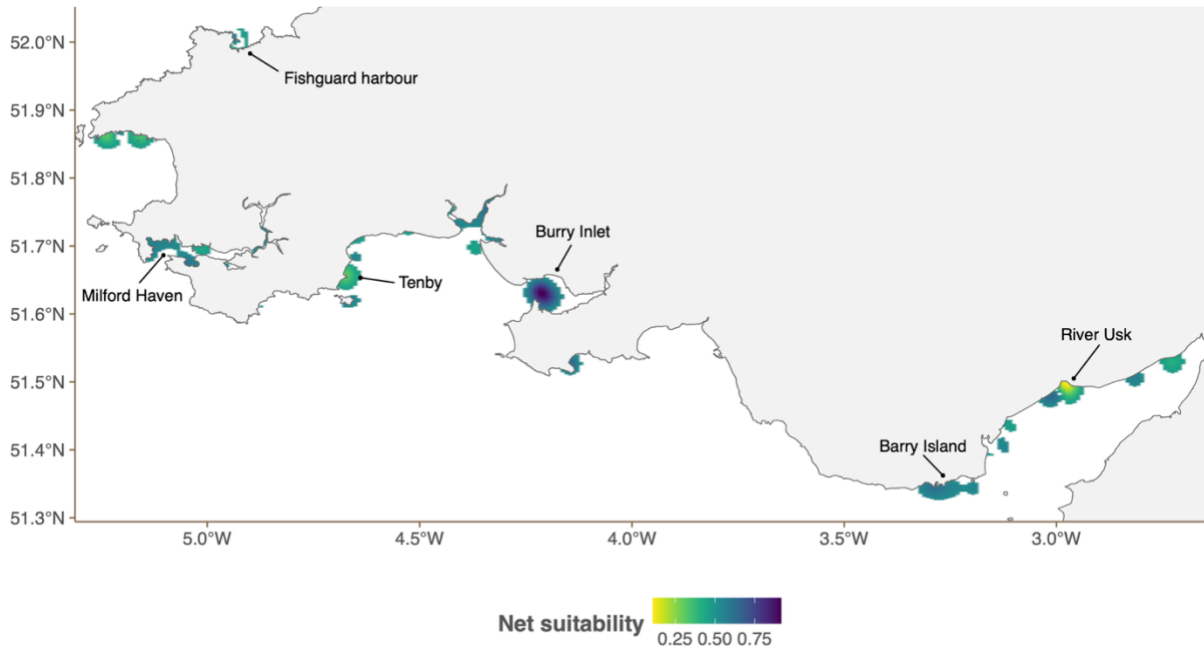
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297 **Results**

298 *Participatory mapping reveals spatially structured restoration opportunity*

299 Local knowledge held by 33 fishers and anglers revealed a structured pattern of perceived
300 opportunity for seagrass restoration across south Wales. The participatory-derived net
301 suitability output showed strong spatial convergence in sheltered coastal bays and
302 estuarine environments, where respondents consistently identified locations favourable for
303 seagrass restoration (Figure 2). Distinct suitability hotspots were evident within the Milford

304 Haven waterway (including Angle Bay, Dale Bay and surrounding inlets), Fishguard Bay, the
305 Burry Inlet, and Barry Island. These areas are characterised by low exposure, shallow
306 subtidal environments, proximate to existing and historic *Z. marina* and *N. noltii* meadows
307 (Unsworth et al., 2026a).



308
309 **Figure 2.** Participatory-derived net suitability for seagrass restoration in south Wales, UK. Kernel density
310 surfaces of perceived suitable and unsuitable restoration locations were generated from stakeholder mapping
311 responses and combined to produce a normalised net suitability output (0 being unsuitable and 1 being
312 suitable).

313

314 In contrast, areas perceived as unsuitable for restoration (despite potential ecological
315 suitability) were strongly associated with zones of high anthropogenic pressure, including
316 port infrastructure in Milford Haven, sedimentary pressure in the River Usk estuary, and
317 heavily visited tourism locations such as Tenby. This spatial contrast of net suitability (0=
318 unsuitable, 1=suitable) indicates that perceptions of restoration potential are shaped not
319 only by environmental conditions but also by patterns of human use and infrastructural
320 constraint.

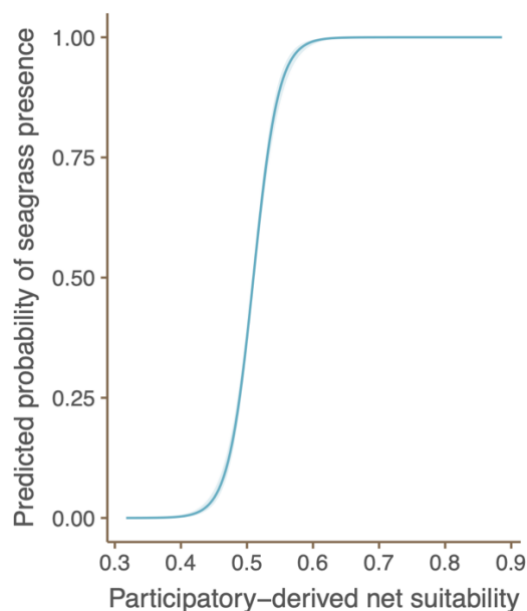
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322 *Participatory-derived suitability conforms with seagrass occurrence*

323 Independent validation using seagrass occurrence data (Jones et al., 2025a) showed strong
324 agreement with participatory outputs. Net suitability derived from participatory mapping

325 was a strong predictor of seagrass presence, with a highly significant positive relationship
326 ($\beta = 52.573 \pm 3.006$ SE, $p < 0.001$). The probability of occurrence increased sharply across
327 the suitability gradient (Figure 3), and model performance was high (AUC = 0.89), indicating
328 strong discrimination between observed seagrass locations and background points.
329 Importantly, this correspondence emerges despite participants being explicitly asked to
330 identify locations suitable for seagrass growth. This distinction is critical as it suggests that
331 respondents are either implicitly conceptualising restoration as the enhancement of
332 historically suitable seagrass environments or identifying areas of existing seagrass. In
333 either case, the strong agreement between perceived restoration suitability and observed
334 distribution indicates that local knowledge is closely coupled to present-day ecological
335 conditions, even when elicited through a future-oriented framing.

336



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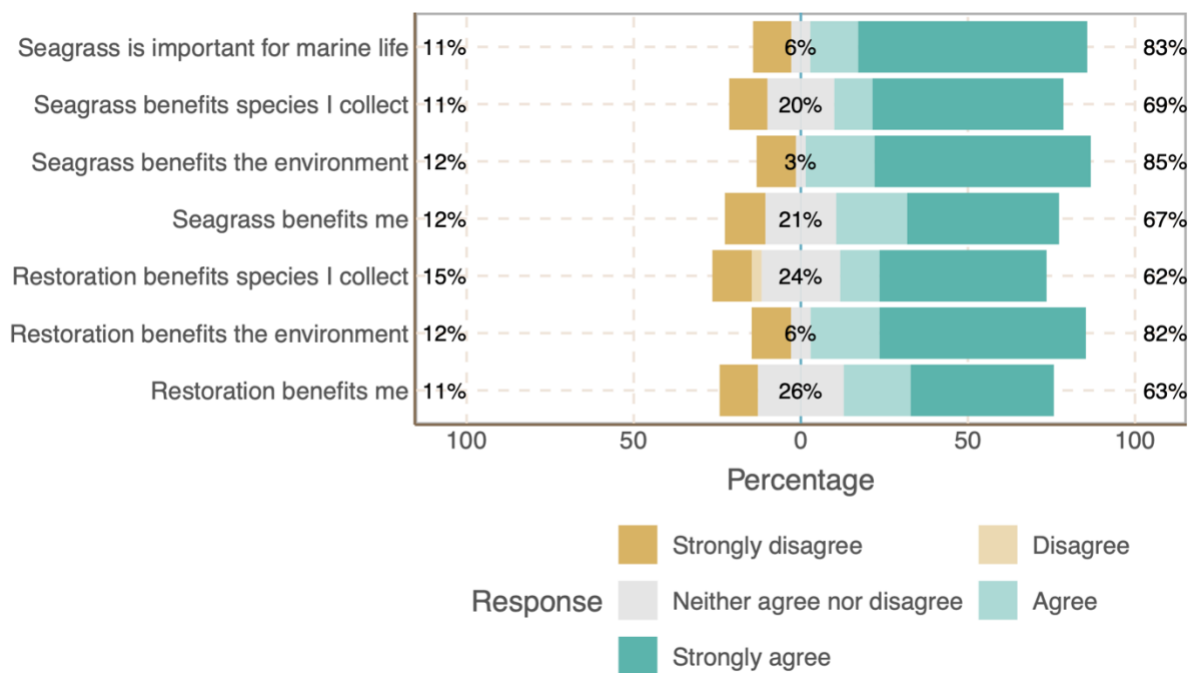
338 **Figure 3.** Predicted probability of seagrass occurrence across a participatory-derived net suitability gradient.
339 The logistic regression curve shows the relationship between participatory-derived net suitability and
340 observed seagrass presence, with the shaded area representing the 95% confidence interval. Higher net
341 suitability values are associated with a greater predicted probability of seagrass presence (AUC = 0.89),
342 indicating that aggregated participatory mapping (Wisdom of Crowds) captured spatial patterns consistent
343 with observed seagrass occurrence.

344

345 *Strong consensus on ecological benefits, with weaker perception of individual gains*

346 Perceptions of seagrass meadows and its restoration potential were strongly positive across
347 south Wales fishers and anglers (Figure 4). The vast majority agreed or strongly agreed that

348 the presence of seagrass benefits the wider environment (85.3%) and marine life (82.9%),
 349 with similarly high agreement that seagrass restoration would benefit the wider environment
 350 (82.4%). Perceived direct benefits to targeted fishery resources were lower, but remained
 351 substantial, with 68.6% agreeing that the presence of seagrass meadows benefits the
 352 species they collect and 61.8% agreeing that restoration would enhance these resources.
 353 Perceived personal benefits were somewhat weaker, with 66.7% agreement for seagrass
 354 presence and 62.9% for seagrass restoration. Across all questions, disagreement remained
 355 low (11–15%), while neutral responses increased for questions relating to individual and
 356 resource-specific benefits (up to 25.7%), indicating greater uncertainty in how ecosystem
 357 recovery translates to direct outcomes for natural resource users.



358
 359 **Figure 4.** Stakeholder perceptions of seagrass ecosystem benefits and restoration outcomes across south
 360 Wales, UK. Diverging stacked bar charts show the distribution of responses (strongly disagree to strongly
 361 agree) for statements relating to (i) the ecological importance of seagrass and (ii) perceived benefits of
 362 seagrass restoration. Bars are centred on the neutral category to illustrate the balance of agreement and
 363 disagreement.

364
 365 *Knowledge base reflects diverse and experienced coastal users*

366 The strength of these spatial and perceptual signals is underpinned by a diverse and
 367 multifunctional knowledge base consistent with a “Wisdom of Crowds” structure (Jones et
 368 al., 2025b). Thirty-four coastal resource users contributed spatial information, representing
 369 a broad range of fishing and harvesting practices. Recreational fishing dominated (76.5%),

370 but substantial overlap with shellfish gathering (41.2%) and bait collection (23.5%) indicated
371 multi-use engagement with coastal systems.

372 Access methods varied, with most respondents operating on foot (79.4%), alongside vessel-
373 based fishing (32.4%) and freediving (26.5%). Nearly one-third reported multiple access
374 modes. Respondents also demonstrated broad ecological knowledge, reporting a wide
375 range of harvested taxa including crabs, lobsters, mussels, oysters, scallops, and cockles.
376 Experience levels were somewhat skewed toward long-term engagement, with 58.8%
377 reporting more than 10 years of experience and only 11.8% reporting less than two years.
378 Spatially, knowledge spanned multiple coastal systems, including Milford Haven, the Burry
379 Inlet, Swansea Bay, Pembrokeshire, and the Severn Estuary, with many respondents
380 operating across multiple regions.

381

382 **Discussion**

383 Site selection is a crucial part of seagrass restoration planning that often defines success
384 when compared with other variables (van Katwijk et al., 2016; Unsworth et al., 2024; Coals
385 et al., 2025; Pansini et al., 2025). Amongst scientists and practitioners, Habitat Suitability
386 Modelling (HSM) has emerged as the de facto way to predict potentially favourable sites,
387 despite limitations (Bertelli et al., 2022). For seagrass systems, these approaches overlook
388 the knowledge held by resource users, despite their long-term engagement with the coastal
389 environments that may be targeted for restoration. This contrasts with suitability models on
390 land, where local knowledge integration is somewhat more commonplace (Stern and
391 Humphries, 2022). Here, we set out to test whether local knowledge held by fishers and
392 anglers could support an alternative approach for seagrass restoration planning.

393 Participatory mapping revealed a coherent spatial structure and with clear areas of
394 consensus for restoration opportunity. In some areas, these participatory suitability
395 “hotspots” aligned closely with known seagrass distribution or areas of historic presence
396 (Unsworth et al., 2026a). Further tests revealed that participatory-derived suitability outputs
397 were strong predictors of independently observed seagrass occurrence.

398 *Restoration is interpreted differently across knowledge systems*

399 An unexpected outcome of this study is that many of the locations identified as suitable for
400 restoration already contain seagrass meadows. This exposes a conceptual mismatch
401 between participatory restoration mapping and assumptions embedded within
402 conventional HSM and applied restoration practice. Conventional restoration planning,
403 particularly when guided by HSM, is often oriented toward novel establishment; identifying

404 new areas where seagrass could be introduced through active interventions such as seeding
405 or transplanting. In contrast, stakeholder responses suggest a framing of restoration that is
406 more closely aligned with enhancement, protection, or recovery of existing systems.

407 This aligns with a broader body of social science literature demonstrating that restoration is
408 a socially constructed and contested practice. While scientific approaches often prioritise
409 recovery towards a reference condition, local and Indigenous perspectives frequently frame
410 restoration as the maintenance or improvement of functioning, lived landscapes (Weng,
411 2015; Robinson et al., 2021). In this context, we suggest that fishers and anglers are not only
412 identifying *empty* space for restoration, but also recognising seagrass as an existing,
413 valuable habitat that may be degraded, vulnerable, or in need of intervention.

414 Rather than identifying new sites for seagrass establishment, the spatial clustering
415 emerging from stakeholder knowledge aligns more closely with an *intervention* perspective,
416 where restoration is directed toward known seagrass locations that may be perceived as
417 degraded, vulnerable, or in need of recovery. This is particularly important in the context of
418 applied restoration methods, which typically refers to active planting of seeds or transplants
419 into bare sediment. The alignment between participatory outputs and existing seagrass
420 distribution potentially suggests that respondents are not only engaging with restoration as
421 habitat creation in new areas, but also as enhancement or recovery of already recognised
422 fishery assets.

423 This divergence highlights a limitation in a direct comparison between participatory
424 suitability outputs and HSMs. While the latter are designed to identify biophysical potential
425 for establishment in unoccupied space, the former may be capturing a management-
426 oriented knowledge system that prioritises familiar, already-vegetated locations where
427 perceived intervention is needed. As a result, the agreement between participatory mapping
428 and observed seagrass distribution should not be interpreted as validation, but as evidence
429 that local knowledge is oriented towards potentially different restoration ontologies focused
430 on improving existing seagrass systems rather than expanding their spatial footprint through
431 new introductions.

432

433 *Implications for restoration planning and site selection*

434 Our findings have direct implications for how restoration planning is structured. Current
435 approaches typically follow a linear sequence involving model-driven site identification, site
436 assessments and restoration trials, followed by stakeholder engagement to resolve
437 conflicts or support implementation, if included at all. We believe this sequence to be
438 suboptimal, particularly for restoration of social-ecological systems like seagrass meadows.

439 We suggest that Incorporating ILK at earlier stages can identify ecologically suitable areas
440 that may be missed or underrepresented in models, highlight fine-scale constraints and
441 pressures (e.g. infrastructure, tourism, localised disturbance) not captured in
442 environmental layers, reveal areas where intervention is socially acceptable or contested
443 and align restoration objectives with stakeholder priorities, improving legitimacy and long-
444 term stewardship.

445 Critically, participatory approaches may also help distinguish between different
446 intervention pathways, including protection of existing meadows, passive recovery through
447 pressure reduction and active restoration in suitable but currently unoccupied areas. This
448 is particularly important given the high cost and variable success of active seagrass
449 restoration. In many systems, improving environmental conditions and reducing pressures
450 may yield far greater returns than planting alone (Ward et al., 2025).

451

452 *Participatory mapping as a social–ecological diagnostic tool*

453 Rather than viewing participatory outputs as direct analogues to HSMs, they may be better
454 understood as social–ecological diagnostic layers. The net suitability output generated here
455 integrates ecological knowledge, historical experience, and perceptions of human pressure
456 into a single spatial product. The distinction between suitability and avoidance is especially
457 informative. While suitable areas aligned with known seagrass habitat, avoidance areas
458 were strongly associated with anthropogenic pressures, including ports, sedimentation,
459 and high-use recreational zones. These patterns highlight constraints that are often poorly
460 represented in HSMs but are critical for restoration success. As such, participatory mapping
461 provides insight not only into where seagrass *could* occur, but where restoration is feasible,
462 acceptable, and likely to persist in the long term.

463 These findings sit within an emerging body of seagrass research showing that local
464 knowledge and participatory approaches can generate spatially explicit, policy-relevant
465 ecological information. Across diverse contexts, these approaches are increasingly used
466 not only to complement conventional mapping, but to strengthen ecological inference and
467 decision-making. For example, combining satellite imagery with stakeholder interviews has
468 revealed long-term seagrass decline trajectories in Brazil and identified locally specific
469 drivers that would have remained undetected using remote sensing alone (Magalhães et al.,
470 2026). In the eastern Aegean Sea, Greece, fisher knowledge has been shown to map
471 *Posidonia oceanica* meadows with high accuracy, even under low-participation conditions,
472 and in some cases outperform official seagrass maps (Alexopoulos et al., 2025).
473 Participatory approaches are also being extended beyond seagrass habitat mapping into

474 applied site selection. In Indonesia, community mapping was used to identify suitable sea
475 cucumber grow-out sites in seagrass-associated systems, integrating both ecological and
476 socio-economic criteria (Ainin et al., 2025). Our study builds directly on this trajectory, but
477 extends it from mapping and habitat assessment into restoration planning.

478

479 *Caveats and future refinement*

480 There are several caveats that should guide interpretation. First, the participatory maps
481 were necessarily presented at broad regional scales. While this was appropriate for
482 identifying spatial patterns and areas of consensus, it was less suited to identifying precise
483 boundaries. Therefore, the points should be interpreted as areas of perceived opportunity
484 or constraint, not as perfect sites. In-person follow-up workshops, targeted interviews, and
485 site-level site assessment would help refine these outputs, identify specific restoration
486 zones, and clarify whether respondents were referring to active planting, passive recovery,
487 protection of existing meadows, or pressure reduction.

488 Participation was voluntary and notably uneven across regions and user groups. Although
489 responses were weighted by region to reduce spatial sampling bias, the sample was still
490 dominated by recreational fishers and anglers, with limited representation from commercial
491 fishers. Because different user groups may hold different knowledge, priorities, and
492 concerns, this does present some bias. For example, the identification of the Burry Inlet as
493 suitable for restoration may not fully capture the perspective of the commercial cockle
494 fishery, highlighting the need for targeted engagement with disenfranchised and
495 underrepresented groups before moving from spatial identification to restoration
496 implementation.

497 Finally, mobile-based mapping introduces positional uncertainty. Some points were likely
498 affected by screen size, zoom level, or user error, and obviously erroneous locations were
499 removed. Future applications could improve precision by combining online mapping with
500 facilitated in-person validation, clearer scale controls, and iterative feedback with local
501 users.

502

503 *Implications for restoration and management*

504 Despite minor caveats, the approach offers clear practical value for seagrass restoration
505 practitioners. Participatory mapping is scalable, relatively low-cost, and suitable for local
506 communities, smaller NGOs, and restoration partnerships that may not have the resources
507 to develop bespoke HSMs. It also brings fishers and anglers into the restoration planning

508 process from the outset, rather than after sites have already been selected. Early
509 involvement can strengthen trust, legitimacy, stewardship, and long-term support, while
510 also helping avoid mistakes caused by selecting sites that are ecologically plausible but
511 socially or operationally difficult. Such mistakes are costly, both for practitioners and
512 funders.

513 We are quick to highlight that participatory mapping should not replace HSMs or field
514 assessment entirely. Instead, it can act as an early social–ecological screening layer to
515 identifying where local knowledge converges, where pressures or conflicts are expected,
516 and where further site-level assessment should be prioritised.

517 To conclude, here we show that fisher and angler knowledge generates spatially coherent
518 and ecologically meaningful information vital for seagrass restoration planning.
519 Participatory-derived suitability did not simply replicate existing HSM but captured different
520 and highly relevant signals that integrating ecological familiarity, human use, perceived
521 pressure, and restoration feasibility. In doing so, it identified both areas of opportunity and
522 areas where intervention may be socially or practically constrained. For seagrass
523 restoration, this latter point is highly relevant. Successful restoration depends not only on
524 whether seagrass can grow, but whether interventions are appropriate, supported, and
525 likely to persist within lived coastal systems. Integrating local knowledge early in restoration
526 planning can therefore improve site selection, guide decisions between protection, passive
527 recovery and active restoration, and build stronger relationships with the communities who
528 know and understand these systems best.

529

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538

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