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**An integrated framework for assessing coastal community
vulnerability across cultures, oceans and scales.**

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1. INTRODUCTION

Across the globe, many coastal communities rely on marine resources for their food security, income and livelihoods and with predicted trends in human populations the number of people reliant on these resources is likely to increase (FAO, 2012). However, the effects of climate change including increased variability are already being experienced by coastal communities and appear to be accelerating (Doney et al., 2012). Depending on a range of factors, including location, these changes are having mild to severe impacts on communities both in direct and indirect ways (Miller et al., 2010). Communities in coastal areas, for instance, are particularly at risk due to sea level rise but also through their dependence on marine resources that are impacted by multiple climate change pressures. A change in the availability and condition of marine resources has consequences on the livelihoods of fishing populations or those who depend directly on fishing as a source of food (Badjeck et al., 2010).

Although mitigating climate change impacts remains the main priority in addressing climate change (IPCC, 2014), it is also important to develop adaptation strategies to climate change in locations where this is possible (Füssell and Klein, 2006; Young et al., 2010), particularly those already experiencing early effects (Hobday et al., 2016; Popova et al., 2016).. While mitigation is generally centred on changing behaviour at the national and/or global level, adaptation is a response that provides affected communities with a locally specific course of action. For a community to develop effective means to adapt to the effects of climate change, it is necessary to first determine and understand the vulnerability of that community (Adger, 2006; Norgaard, 2011). Once the vulnerability of a community is well understood, appropriate adaptation options can be developed through collaboration between local stakeholders, researchers and managers, and put into action through government and non-government

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3 24 institutions (Savacool et al., 2015). Recognising these dual stages (assess vulnerability and
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5 25 develop adaptation options) helps avoid a key criticism of vulnerability studies, in that they only
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8 26 determine the vulnerability of certain communities without providing the pathways and means to
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10 27 address the issues uncovered (Preston, 2012). It also reduces the chance of developing adaptation
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12 28 strategies that are ill-suited to the often complex social and ecological systems in which they are
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15 29 to be implemented.

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17 30 Vulnerability studies, which have been used to address a range of research aims and have
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19 31 used a diversity of approaches, have attracted considerable scrutiny. The debate has centred
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21
22 32 around: measurement of an often imprecise or poorly defined concept (Füssell, 2007); delivery
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24 33 of a vulnerability score for a community that is relatively meaningless (Smit and Wandel, 2006);
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27 34 the aggregation of indicators to develop scores which can mask or overlook important factors
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29 35 that heighten or reduce vulnerability (Preston, 2012); assessments made at only one scale and
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31 36 then up- or down-scaled to make comparisons at other levels of complexity (Cutter et al., 2003),
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34 37 and; the high confidence placed in vulnerability mapping that is often performed using low-
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36 38 resolution secondary data (Preston et al., 2011). Often these studies leave policy makers and the
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38 39 communities that they serve unsure what these vulnerability results mean for their future
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41 40 livelihoods and overall sustainability, and what responses (if any) are implied.

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44 41 This paper provides an improved framework for assessing the vulnerability of coastal
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46 42 communities across cultures, oceans, and scales, and suggests ways in which adaptation
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48 43 strategies can be conceptualized and implemented more effectively. In this context vulnerability
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50 44 integrates the qualities of being exposed to and sensitive to change in the marine environment
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52 45 and the degree to which adaptation strategies can counteract this. First, we describe an integrated
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55 46 vulnerability framework developed by members of the Belmont funded project known as

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3 47 GULLS (Global Understanding and Learning for Local Solutions), emphasising its strengths in
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5 48 addressing key issues of cross country vulnerability comparisons and scaling up, and in
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7 49 providing a basis for integrated social and ecological modelling of climate change adaptation in
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9 50 the broader GULLS project. Second, we describe the unique manner in which the conceptual
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11 51 framework has been translated into a vulnerability analysis (through various ethnographic
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13 52 methods) and implemented in marine-based coastal communities across southern hemisphere
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15 53 countries, namely Australia, Brazil, India, South Africa, Madagascar, and Solomon Islands (Fig.
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17 54 1). Finally, we provide a summary of results at the *country* level to evaluate the appropriateness
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19 55 of the new integrated framework and show its relevance and applicability. This framework was
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21 56 developed as part of a Belmont Forum multilateral-funded project (initiated in 2013 by the
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23 57 authors and various other colleagues [Hobday et al., 2016]) with the overall aim of addressing
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25 58 the issue of coastal community vulnerability more comprehensively and providing meaningful
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27 59 adaptation strategies to both policy makers and local populations in regions with relatively high
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29 60 exposure to climate driven changes in the marine environment.
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36 **FIGURE 1 GOES HERE**

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39 62 A unique aspects of the GULLS project is that it focuses on marine-dependent coastal
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41 63 communities in a number of *hotspot* countries, where coastal seas are warming faster than in
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43 64 other nations' marine areas. These marine hotspots are seen as priority areas for research as they
44
45 65 are places where the effects of warming oceans are being observed and experienced first
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47 66 (Hobday and Pecl, 2014). These hotspot areas are in essence natural laboratories for biological
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49 67 and social change, and they provide valuable case studies for identifying generic and scalable
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51 68 measures and pathways of adaptation to the likely impacts of climate change for other coastal
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53 69 communities of warming seas in the near future (Pecl et al., 2014a). Distilling globally relevant
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3 70 learning outcomes from the GULLS case study countries is imperative, as many other locations
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5 71 may not have the capacity to carry out the level of preliminary research carried out as part of the
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7 72 GULLS project, due to lack of funding, expertise, political resolve or where the critical time
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9 73 frame for action requires immediacy.

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13 74 A key component of the GULLS project was to collect rich, local level, social
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15 75 vulnerability data which would provide a fine understanding of the local scale processes
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17 76 influencing communities' vulnerabilities while allowing for the data to be scaled up to a
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19 77 regional, country, and global levels allowing integration with ecological and oceanographic
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21 78 models and comparisons among hotspot communities and countries. The vulnerability
22
23 79 comparisons at the different scales, combined with the relevant ecological and oceanographic
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25 80 predictions, will provide accelerated learning mechanisms for communities likely to experience
26
27 81 similar stressors and changes to their way of life in the future (Hobday and Pecl, 2014). Gaining
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29 82 new insights into marine social and ecological systems using different ecological modelling
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31 83 approaches combined with scalable social, economic, cultural and governance vulnerabilities
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33 84 will ultimately add to complex systems science (Berkes, 2006) and better prepare us for
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35 85 management of these systems in the Anthropocene. Conducting vulnerability assessments in this
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37 86 complex, multi-scale, cross-cultural context required development of a new conceptual
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39 87 vulnerability framework and implementation approach.
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48 89 **2. Existing Vulnerability Frameworks**

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51 90 There are a multitude of vulnerability frameworks reflecting different disciplinary backgrounds
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53 91 of vulnerability analysts and with different aims and objectives that are already available and in
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55 92 use (Cutter et al., 2003, Eakin et al., 2009; O'Brien et al., 2004; Adger et al., 2009). Each
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3 93 framework comprises different components including, for instance, risk-hazard models within
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5 94 geographical studies (Karim and Mimura, 2008), pressure-release models (Schröter et al., 2005)
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7
8 95 and social vulnerability/adaptive capacity models (Cutter et al., 2003; Vincent, 2004). The
9
10 96 Intergovernmental Panel on Climate Change (IPCC, 2001) vulnerability framework has been
11
12 97 utilised in many different studies as it comprises both ecological and social components and can
13
14
15 98 be applied at a variety of scales from global to local. One criticism of using this framework
16
17 99 alone centres on the potential to simplify adaptive capacity to economic components (Table 1).
18
19 100 Other studies have tried to embed more nuanced social and economic components within the
20
21 101 frameworks they have used. Allison and Horemans' (2006) study linked the Sustainable
22
23 102 Livelihoods Approach (SLA) with the Livelihood Vulnerability Index (LVI) and had a strong
24
25 103 emphasis on policies and institutions. However, for studies interested in the impacts of climate-
26
27 104 induced change, this framework lacks any means to integrate climatic exposure and adaptation
28
29 105 strategies. Another vulnerability framework is the Food Security (FS) framework (Table 1)
30
31 106 which focuses on food availability and access. In isolation this framework often lacks meaning at
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33 107 the local level, and as such it tends to be used at higher (coarser) scales for regional and national
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35 108 decision making.
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41 109 A number of studies have responded to the multi-dimensional nature of vulnerability in
42
43 110 complex human-environment contexts by combining, or integrating, multiple frameworks. In the
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45 111 marine context, for example, Himes-Cornell and Kasperski (2015) developed an integrated
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47 112 vulnerability framework for analysing Alaska fishing communities that considered exposure
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49 113 through a rapidly changing Alaskan environment, rapid local resource dependence changes, and
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51 114 community adaptive capacity to climate change. Similarly, Colburn *et al.* (2016) developed a
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53 115 multi-dimensional framework to measure the vulnerability of US East coast fishing communities
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3 116 by analysing a series of new indicator under the context of climate change, and building on the
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5 117 National Oceanic and Atmospheric Administration's (NOAAs) existing Community Social
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7 118 Vulnerability Indicators (CSVIs). Allison et al. (2009) used the IPCC approach to estimate the
8
9 119 vulnerability of national economies to climate-induced changes in the marine environment at the
10
11 120 global scale. Metcalf et al. (2015) applied a similar approach to three geographically dispersed
12
13 121 case studies in Australia. An integrated LVI and IPCC approach was used to estimate social
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15 122 vulnerability in a single region of Mozambique (Hahn et al., 2009) and a study of a specific
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17 123 urchin fishery in the USA (Chen et al., 2014). Cinner et al. (2012) also used an integrated
18
19 124 IPCC/LVI approach in a regional comparison of vulnerability to climate change of communities
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21 125 in the Indian Ocean. Most recently, Mohan and Sinha (2015) combined IPCC and LVI
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23 126 frameworks to assess vulnerability to climate change in the Ganges River basin. However, to the
24
25 127 best of our knowledge there has been no attempt made to develop an integrated approach to
26
27 128 compare the vulnerability of coastal communities situated in different ocean basins characterised
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29 129 by different general climate norms, yet now all affected by warming oceans disproportionately
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31 130 (vis-à-vis other countries), and across countries with very different social, cultural, and
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33 131 development levels.
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40 132 **TABLE 1 GOES HERE**
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44 134 **3. An Improved Integrated Vulnerability Framework**

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46 135 The vulnerability framework presented here (Fig. 2) comprises two high level components
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48 136 representing the biological and human subsystems. The framework allows for scaling up of the
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50 137 human vulnerability analysis to allow integration with change information available for the
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52 138 ecological system. The level of environmental exposure combined with the biological sensitivity
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54 139 of different marine species determines the ecological vulnerability in the ecological subsystem
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3 140 (Pecl et al., 2014b). The ecological vulnerability, in turn, has a direct influence on the socio-
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5 141 economic subsystem. For instance, a crustacean species may be biologically very sensitive to
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7 142 warming ocean temperatures (see for instance Pecl et al., 2014b) and at its biological limit in a
8
9 143 particular fast warming hotspot (thus making it ecologically vulnerable). In addition if the
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11 144 crustacean species is economically important for the local commercial fishery then the potential
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13 145 impact (in the human system) of climate driven change in the marine environment for this
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15 146 species will be relatively high.
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20 147 **FIGURE 2 GOES HERE**
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23 148 The potential impact in the socio-economic subsystem, aside from being influenced
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25 149 directly by the ecological characteristics of the system, is also defined by the dependence on
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27 150 marine resources of the people making up the socio-economic subsystem. Therefore resource
28
29 151 dependence is part of the sensitivity component of the vulnerability framework which is not only
30
31 152 defined in terms of economic dependence (often considered in isolation in developed countries
32
33 153 and in many vulnerability assessments) but importantly also considers the level of social,
34
35 154 historical and cultural dependence (which may be of importance in developing countries or
36
37 155 where indigenous marine uses are relevant). Together, in the human system, exposure, resource
38
39 156 dependence and adaptive capacity impact socio-ecological vulnerability (Metcalf et al., 2015).
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41 157 To provide a robust methodological approach for measuring vulnerability of marine dependent
42
43 158 coastal communities to climate change that was applicable in countries with various levels of
44
45 159 economic development, the sustainable livelihoods approach (SLA) forms the core element of
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47 160 adaptive capacity. An additional component of flexibility was incorporated to further refine the
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49 161 assessment of adaptive capacity (Fig. 2). The degree of flexibility across multiple scales
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51 162 (personal, occupational and institutional) through to institutional was included thereby better
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3 163 measuring people's and institutions' potential to influence their current situation and adapt to
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5 164 changing future conditions (Marshall and Marshall, 2007; Marshall, 2010). The proposed
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7 165 framework was not developed to directly assess risk (i.e., the possibility that an action or activity
8
9 166 will lead to a loss or undesired outcome). However, risk is a component of the interactions of
10
11 167 vulnerability, exposure and hazards (Oppenheimer et al., 2014) because vulnerability changes the
12
13 168 probability that a risk will lead to undesirable outcome. Rather, the exposure to natural hazards
14
15 169 was assessed in terms of household exposure (to storms, floods, droughts and shoreline changes),
16
17 170 and it is thus also possible to evaluate risk using the proposed framework. The importance of
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19 171 considering vulnerability and adaptation together in the integrated framework is shown in Figure
20
21 172 2 with the development of adaptation options being informed by and in turn influencing both
22
23 173 ecological and socio economic components.

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29 174 Individual components within the exposure, sensitivity and adaptive capacity categories
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31 175 of the socio-economic subsystem were then further expanded to provide more detailed
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33 176 descriptors, or sub-components and measurable indicators (Fig. 3). The detailing of the sub-
34
35 177 components allowed the team to develop a generalised survey instrument with individual
36
37 178 questions. This research instrument provided information on each indicator, thereby allowing us
38
39 179 to map the different components of the integrated framework (Fig. 3). The final framework
40
41 180 comprised a total of 90 subcomponents, with 255 indicators providing the link between the
42
43 181 conceptual framework (Fig. 2) and the survey methods that were developed and used across
44
45 182 hotspot locations (allowing for comparison among hotspots at these higher levels). A link to the
46
47 183 full survey can be found at <http://gullsweb.noc.ac.uk/communitysurvey.php>. Note that this
48
49 184 conceptual framework not only emerges from the experience of GULLS members during this
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51 185 research but also from their prior work with issues surrounding coastal communities and
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3 186 vulnerability (e.g., Gasalla and Diegues, 2011; Shyam et al., 2014; van Putten et al., 2015;
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5 187 Aswani et al. 2015).

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8 188 **FIGURE 3 GOES HERE**
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12 13 190 **4. METHODS**

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15 191 Comparative research into vulnerability is a crucial guide for resource allocation and policy, both
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17 192 at a national level and to aid international donor organizations (Vincent 2007). But since
18
19 193 vulnerability assessments often remain ad hoc and use many indices, comparisons across
20
21 194 countries and regions can therefore only be made with caution, especially since (due to the
22
23 195 relative newness of vulnerability assessments using indices for adaptive capacity) researchers
24
25 196 cannot yet be certain of the reliability of their chosen indicators to measure complex local
26
27 197 realities (Vincent 2007). To overcome such uncertainty, researchers have to be transparent with
28
29 198 their methodology. Also, it is important to move beyond offering a mere comparative snapshot of
30
31 199 different communities' relative vulnerability and to expose the underlying dynamics of what
32
33 200 constitutes this vulnerability (Thomalla et al. 2006). This entails the use of sophisticated
34
35 201 statistical models in order to prevent data loss through aggregation. Such techniques in
36
37 202 vulnerability studies were first introduced in the seminal paper by (Cutter et al. 2003) and later
38
39 203 expanded to use non-parametric techniques (Hahn et al. 2009). However, to obtain truly
40
41 204 comparative results vulnerability assessments must use the same indicators as different
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43 205 methodologies have been shown to generate significantly different results (Yoon 2012, Wirehn
44
45 206 2015). The necessity for broad-based vulnerability assessment (Bennett 2016) results in a
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47 207 proliferation of vulnerability indicators, each increasing the uncertainty that the indicators in
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49 208 question possess construct validity (Vincent 2007). As detailed below, to ensure that the data
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3 209 gathered would be truly comparative, the proposed framework and the survey instrument was
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5 210 constructed through careful collaboration and based upon best practice gleaned from the
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8 211 literature.

10 212 **4.1. Producing a Survey**

13 213 The first step to develop a survey methodology to establish the vulnerability of the coastal
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15 214 communities in the hotspot countries was to conduct a literature review and account for the
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17
18 215 weaknesses of typical vulnerability studies expressed in the literature (e.g., Preston, 2012;
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20 216 Preston et al., 2011; Füssel, 2007; Smit and Wandel, 2006; Cutter et al., 2003). In particular, to
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22
23 217 measure social vulnerability in different countries, the approach needed to be sensitive to local
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25 218 cultures and social contexts, both in terms of the process and the method used to gather the
26
27 219 necessary information. Ultimately, the information gathered would need to be comparable
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29
30 220 between hotspots and the approach implementable across other (non-GULLS) hotspot and non-
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32 221 hotspot countries. The information needed to be generally comparable between locations, yet
33
34 222 specific enough to make it possible to take into account the local context to identify ways and
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36
37 223 means for communities to adapt to their potentially common vulnerabilities or for different
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39 224 communities to learn from their unique differences. The methods developed as part of the
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41 225 GULLS project, therefore, were also designed to allow the social vulnerability analyses to be
42
43 226 integrated with the GULLS ecological and oceanographic prediction and vulnerability research
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45
46 227 (reported in Hobday et al., 2016). Detailed primary data was collected for the social vulnerability
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48 228 analysis to ensure nuanced layers of human systems data fitted with the use of secondary data (or
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50 229 previously collected primary data) for the ecological and oceanographic modelling (Preston,
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52
53 230 2012).

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3 231 The second step was to develop a survey instrument and to field-test it in most
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5 232 participating countries. Field testing was carried out for approximately two weeks in each
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7 233 country and questions that did not produce reliable data were identified and subsequently
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9
10 234 improved or omitted. For example, during testing, the rating questions in the survey which use a
11
12 235 Likert scale (with the typical strongly agree to strongly disagree continuum) were found to be
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14 236 difficult to interpret or answer by many participants. Rating questions were changed to item-
15
16 237 specific responses to allow for clearer comprehension and in turn greater accuracy and reliability
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18 238 of the data collected (Saris et al., 2010). As an example, a question asked in testing was “Do you
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20 239 feel that you belong to this community?” and the respondent was asked to choose from “Strongly
21
22 240 agree”, “Agree”, “Disagree”, “Strongly disagree”. Many respondents answered “Yes” or “No”
23
24 241 despite the question format. This question was modified to give respondents a choice of
25
26 242 “Strongly belong”, “Belong”, “Do not belong”, “Do not belong at all”. In sum, this pre-screening
27
28 243 served to modify and streamline the survey instrument and allowed the individual country
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30 244 project teams to adjust questions to account for local circumstance, language, and
31
32 245 understandings. Overall, more than 80% of the final survey questions were exactly the same
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34 246 across all hotspots indicating a flexible but, more importantly, a transferable method. In the less
35
36 247 developed countries, for instance, an understanding of adaptive capacity was gained through
37
38 248 questions about availability of electricity, water and sewage. These questions were not applicable
39
40 249 in Australia, for instance, as the vast majority of people have access to these services. Overall,
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42 250 local level surveys in each hotspot country included at least one question from each sub category
43
44 251 to ensure data spread across all domains of the framework (i.e., data on all of the subcomponents
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46 252 and indicators) and to ultimately ensure cross- country comparability (Fig. 4).

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54 253 **FIGURE 4 GOES HERE**
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5 255 Due to some further country differences including diverse resources and research person
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7 256 power availability, receptiveness of the target audience (due to survey fatigue), literacy rates, and
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9 257 access to online resources among the country hotspots, two different survey methods were
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11 258 applied (face-to-face and online surveys). The differences in sample size and proportion,
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13 259 research team size, field resources and interviewer training, language and use of translator
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15 260 services across the hotspot countries is shown in Table 2. For instance, the Australian researcher
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17 261 team conducted an online survey rather than a field-based household survey as adopted by the
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19 262 other countries. The web surveys were however unable to collect the large amounts of additional
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21 263 contextual data provided by informants in the field-based surveys. In South Africa, a single
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23 264 translator was used to conduct the household surveys, several translators were used in India, two
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25 265 in Madagascar, and four translators were used to perform the surveys in the Solomon Islands,
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27 266 whereas in Brazil and Australia *in situ* translators were not required. All translators were trained
28
29 267 to use and carry out the survey before the data collection began. By developing the common
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31 268 framework and survey these differences could be accommodated, thereby ensuring the validity
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33 269 of future cross country comparison and analysis. This was challenging given that most countries
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35 270 involved in the research have both local and international researchers, spanning a wide range of
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37 271 academic experience and disciplinary backgrounds, yet this diversity of ideas, knowledge,
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39 272 networks and expertise enriched the research and discussions.
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47 273 **TABLE 2 GOES HERE**

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49 274 The sampling strategy consisted in ensuring comparability of estimated vulnerability
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51 275 across countries and agreeing in a uniform survey sampling universe (i.e. the groups/people to be
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53 276 sampled within the study community). The sampling universe was stratified to encompass
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3 277 randomly selected “people with regular interaction with the ocean”. This meant the survey
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5 278 would not be limited to fishers (artisanal, customary, commercial or recreational) but could
6
7 279 include anyone who interacted regularly with the sea. Note that industrial fishing was not
8
9 280 included in our sampling universe as this would limit the comparability across all countries and
10
11 281 sites. The unit of analysis for both the face-to-face and online surveys was the household
12
13 282 randomly selected from a suitable and representative pool both in the pre-testing and in the final
14
15 283 survey implementation. Household level surveys allowed detailed levels of information to be
16
17 284 collected but was not as intensive as working at the individual level, especially when undertaken
18
19 285 face-to-face. The target survey sample was made up of either members of a physical location (i.e.
20
21 286 a coastal community or town) or members of a stakeholder group. Coastal communities were
22
23 287 selected based on the following set of criteria: sites had to be “small” communities (<5000
24
25 288 inhabitants) and be marine dependent communities. To obtain a representative sample, the
26
27 289 number of people interviewed differed between the two target audience groups due to their
28
29 290 absolute size. For example, a stakeholder group can consist of, for example, only 50 members
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31 291 where a coastal community can contain a total of 5000 members. Survey samples in each coastal
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33 292 community were chosen to be representative (demographically) of the people who lived in that
34
35 293 community. In the case of the survey sample for stakeholder groups (as in Australia where
36
37 294 recreational fishers were targeted), the sample had to be representative of the total recreational
38
39 295 fisher population.

46
47 296 In Brazil face-to-face surveys were conducted *in situ* in households of eight different
48
49 297 coastal communities from the South Brazil Bight (SBB) coastal zone. The SBB corresponds to
50
51 298 the most industrialized and urbanized region of the country but still shelters several traditional
52
53 299 fishing communities. The selected sites represented a comprehensive sample of fishing

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3 300 communities, exhibiting a suite of different characteristics within the SBB. A total of 151
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5 301 households were surveyed across the eight selected communities. Additional fishers' perceptions
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7 302 of climate and ocean conditions were also conducted based on the ethno-oceanographic
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9 303 framework (Gasalla and Diegues, 2011) and formed the basis for understanding exposure
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11 304 (Martins and Gasalla, 2017).
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15 305 In India the research team engaged with local people from the respective communities
16
17 306 (mostly educated and committed women and proactive college students) to conduct the survey.
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19 307 A pilot testing of the survey was carried out initially by the research team of the CMFRI
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21 308 institute, after which appropriate country specific modifications were made to the survey
22
23 309 questions. First, the team developed relationships and rapport with the local self-government
24
25 310 officials (Panchayath), line departments and women self-help groups within the communities by
26
27 311 regular visits and focussed group discussions. The project inception took place in the village,
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29 312 which ensured active community participation in the survey process and acceptance of the
30
31 313 project from the beginning. Second, each self-governed local district involved in the study
32
33 314 educated local people for further training, prior to the implementation of survey. Third, the
34
35 315 selected people were trained in topics covering climate change, vulnerability, sensitivity,
36
37 316 exposure, adaptive capacity and resource management. They were also specifically trained in
38
39 317 conducting household surveys among fishers. A total of 800 households were surveyed in the
40
41 318 study across a number of communities.
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48 319 Five different communities within the same coastal region of the southern coast of South
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50 320 Africa were chosen for the study. An important fishery in this area is the handline fishery
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52 321 performed by small crews of fishers on boats that leave from small harbours or river mouths. The
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54 322 sites consist of a spectrum of different size communities, proportion of households with regular
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3 323 interaction with the ocean and remoteness. The social vulnerability surveys were carried out by
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5 324 one researcher and a translator, as most communities speak Afrikaans. Extensive training of the
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8 325 translator was performed initially and then pre-testing of the survey was carried out to ensure
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10 326 accuracy of the translated survey in the local dialect as well as optimal understanding by survey
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12 327 participants. Overall, 65 surveys were conducted and following on from these, group focus
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15 328 sessions were carried out in each community to feed back some of the main early findings of the
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17 329 surveys.

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20 330 In Australia, an online survey was developed to gather the vulnerability information
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22 331 because face to face interviewing was very difficult logistically (as people are spread out over a
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24 332 large region). The survey was applied to the recreational fishing community. The respondents
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27 333 were made aware of the survey using a social media site (Redmap) and a small incentive prize
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29 334 was offered for participation in the survey. The survey was aimed at people engaged with the
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31 335 ocean (as per the sampling protocol) but the large majority were recreational fishers in coastal
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34 336 communities of the south east of Australia which is the area that corresponds to the marine
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36 337 climate change hotspot area. A total of 56 useable surveys were obtained in this manner. A
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38 338 second call for recreational fisher participation was sent out to recreational fishers after the
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41 339 attempt to engage and survey commercial fishers was unsuccessful. The commercial fishing
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43 340 industry is the subject of many surveys in Australia and survey fatigue has become a serious
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45 341 problem for researchers. Even though engagement with the fishing sector occurs through
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47 342 government departments, engagement by research organisations and individual researchers often
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50 343 results in low response rates. Nevertheless, the recreational fishing sector in Australia takes a
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52 344 considerable portion of the catch, and participation is high in all States and Territories, at over
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54 345 19.5% of the Australian population (Henry and Lyle, 2003)

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3 346 The Solomon Islands, a non-hotspot country, was included in our study as a means of
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5 347 ascertaining the extent to which the GULLS social vulnerability survey could be applied in other
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7 348 countries. No pre-testing was done for the Solomon Islands region due to the high costs
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10 349 involved in travelling to the area, while complications such as a lack of printing facilities
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12 350 prevented the modification of the questionnaire on-site. However, a researcher with 27 years'
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14 351 work experience in the area and a complete grasp of two local languages modified the
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16 352 questionnaire to suit the specific environment. The second researcher received training in
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18 353 conducting the GULLS household survey from members of the South African research team in
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20 354 their field site prior to surveying in the Solomon Islands, which improved the reliability of the
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22 355 data gathered. Four educated translators with a grasp of the English language were used.
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25 356 Translators were trained to understand the point addressed by each survey question and agreed
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27 357 on the wording they would use when translating questions from English into the local language
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29 358 *in situ*. The response was then again recorded in English. A total of 110 surveys were completed
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31 359 for this region. Finally, Madagascar was surveyed by the same two researchers who performed
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33 360 the Solomon Islands survey. Pre-testing was again not possible due to the logistical
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35 361 complications of working in a remote location. The original survey text was first translated into
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37 362 French by two local students who then presented the questions in Malagasy to respondents, with
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39 363 the response recorded in English. A total of 48 surveys were completed between the two
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41 364 communities.
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47 365 The presented framework was utilized for cross-cultural comparisons among the different
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49 366 coastal communities in the different hotspot countries. Not only did we make comparisons within
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51 367 each hotspot country across their different study sites but we identified interesting differences
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53 368 and/or similarities in vulnerability and adaptation patterns in countries where coastal
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3 369 communities have no clear contact or links. Including the Solomon Islands in this first
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5 370 application of the integrated framework provides an early indication of the transferability of the
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8 371 application of the survey method to non-hotspot countries. This created the potential for
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10 372 improved learning and adoption of effective adaptation measures from different places. In order
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12 373 to perform the statistical analysis, several questions were asked including: do commonalities
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14 374 exist across the chosen hotspots and do they manifest in similar ways? The survey approach
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16 375 developed in this paper allowed researchers to address the question as to what makes some
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18 376 communities more vulnerable than others to environmental change. The model's inherent
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21 377 flexibility while mapping back to a robust core framework allowed for quantitative and
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23 378 qualitative data to be collected, analysed, and compared within and among countries and regions.
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27 379 **4.2. Statistical Analysis**

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30 380 The total number of observation for all countries combined was 1,276 but accounting for
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32 381 incomplete observations, a dataset of 1,237 observations was retained. We restrict our survey
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34 382 data analysis to rating questions only and the variables for the current analysis are therefore
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36 383 *ordered ordinal*. For example, the survey question 'How difficult has it become to catch fish in
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38 384 the areas you fish?' has four levels: not difficult at all, not very difficult, somewhat difficult, and
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40 385 very difficult. The value of the rating question was between 1 and 4, with one being the better
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42 386 outcome and four being the worst outcome. Because some of the rating questions had 3, 5 or 6
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44 387 possible categories, these were normalised to a value between 1 and 4 to allow for interpretive
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46 388 consistency. We report the results based on the average and median score for each of the
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48 389 components (we test for consistency and difference between the result using average and median
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50 390 scores). Our dataset has three characteristics which guide our use of statistical tools: the data is
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52 391 ordered categorical, not normally distributed, and the sample sizes are unequal (i.e. India's
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3 392 sample is larger than that of other countries). To determine statistical differences between
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5 393 countries we use nonparametric tests/distribution-free tests to account for a non-normal
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7 394 distribution and unequal sample sizes. In a nonparametric test the null hypothesis is that the two
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9 395 populations are equal, which is interpreted as the two populations are equal in terms of their
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11 396 central tendency. The test used to establish statistical differences between countries allows for
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15 397 unequal sample sizes.

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17 398 We generated explanatory statistics such as samples means, medians, and standard
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19 399 deviation and test the data for correlation for all questions and within vulnerability categories.
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21 400 The Kruskal-Wallis Test was used to determine whether there are significant differences between
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23 401 hotspot countries with respect to rating the components for sensitivity, adaptive capacity, and
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25 402 exposure. Using the Kruskal-Wallis Test, we determined whether the population distributions are
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27 403 identical *without* assuming a normal distribution. The relationship between the vulnerability
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29 404 scores and the three categories can be visualised as indicated in Figure 5. For instance, if
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31 405 sensitivity or exposure are reduced (indicated by the inward facing arrows in b) in Figure 5 then
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33 406 vulnerability can be reduced. Alternatively, if adaptive capacity can be increased this will reduce
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35 407 overall vulnerability to change in the marine environment (indicated by the outward facing
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37 408 arrows in c, Fig. 5). Finally, we analyse country results in the context of socio-demographic
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39 409 characteristics (obtained from publically available dataset) to determine the relationship between
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41 410 the vulnerability scores and these social, economic, and demographic indicators and to gauge if
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43 411 country scores were significantly different. The following results are only an overview of our
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45 412 data at the *country level* to illustrate the applicability of this approach, as a more detailed analysis
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47 413 is presented elsewhere (van Putten et al. unpublished data).

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55 414 **FIGURE 5 GOES HERE**

415 5. RESULTS

416 The countries included in this analysis are experiencing rapid change in the marine environment
417 and their empirically derived sensitivity, adaptive capacity, and exposure gives insight into their
418 relative vulnerability. Based on 80 survey rating questions applied in six different countries,
419 coastal communities in Madagascar followed by India and South Africa are most vulnerable to
420 change in the marine environment. Overall the difference in vulnerability to change in the marine
421 environment between countries is statistically significant (Kruskal-Wallis chi-squared = 299.69,
422 degrees of freedom = 5, p-value < 2.2e-16) (Fig. 6). The relatively high vulnerability of
423 Madagascar is mainly attributable to Madagascar's economic dependence on marine resources
424 making it very sensitive to change in the marine environment. Madagascar has high exposure
425 (Fig 7a) caused by the number of observed changes and unfavourable perceptions and attitudes.
426 The low adaptive capacity of coastal communities in Madagascar (Fig 7b) is caused by the low
427 personal and occupational flexibility and limited physical capital. India is also vulnerable to
428 change in the marine environment due to the high exposure (Fig 7a) that is mainly attributed to
429 the high shoreline change and susceptibility to flood. The exposure of Indians coastal
430 communities is not compensated for the relative low sensitivity (Fig 7c) attributed to the low
431 attachment to the fishing occupation and intermediate adaptive capacity. The low vulnerability in
432 Brazil is attributed to a high adaptive capacity and low exposure which compensates for a
433 relatively high sensitivity due to a strong attachment to fishing and attachment to place. Several
434 country characteristics were found to be highly correlated to the level of vulnerability.
435 Government effectiveness was most highly correlated to country level vulnerability, indicating
436 that a less effective government and high vulnerability tend to go together. Even though country
437 characteristics are found to be related to the empirically derived vulnerability scores, it does not

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3 438 explain the highly complex relationship between coastal community sensitivity, adaptive
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5 439 capacity, and exposure to change in the marine environment that ultimately underpins their
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7 440 vulnerability (Table 3). Our framework, nevertheless, can down-scale to analyze community
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9 441 level sensitivity, adaptive capacity, and exposure and produce cross-community comparisons at
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11 442 the national level or cross-culturally.
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15 443 **FIGURE 6 GOES HERE**

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18 444 **FIGURE 7 GOES HERE**
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24 446 **6. DISCUSSION**

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26 447 Each of the examined hotspots is connected by the changes projected in the marine environment
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28 448 as shown by the general analysis. For further analysis, however, the hotspot communities
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30 449 analyzed in different countries have very different socio-economic and environmental
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32 450 characteristics. Therefore, in future analysis collapsing variables from the household survey into
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34 451 a small number of indicators or domains will need to be context specific. It is acknowledged that
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36 452 there are dangers in integrating local level indicators from such vulnerability surveys and
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38 453 aggregating data to allow for comparisons at higher levels. Yet aggregation is necessary to
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40 454 enable the rich local level data to provide further insight than only for a few selected
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42 455 communities. The proposed framework is flexible for the local contexts while maintaining
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44 456 important information that is consistent across all sites to allow for comparisons. The framework
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46 457 should be used as a tool to calibrate and validate regional to national level vulnerability
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48 458 assessments. The framework developed here gives us an opportunity to compare how the
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50 459 vulnerability assessment differs when estimated at different scales.
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3 460 The framework can be used to merge a series of secondary datasets identified as indicator
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5 461 variables from widely available global data such as climate change projections, population
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7 462 projections, gross domestic production (GDP) and infrastructure data. Each component of the
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9 463 vulnerability framework can be aggregated from the gridded level to the hotspot and compared to
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11 464 the same components estimated from household surveys. This enables the research to begin
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13 465 considering the bias in vulnerability analyses performed on secondary data at the global level.
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15 466 Building on the survey the scaling up of results can be achieved through modelling and climate
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17 467 change scenario development and socio-demographic changes like population growth. This
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19 468 scaling up component creates a global gridded surface of marine and terrestrial exposure to
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21 469 climate change. The indicator variables selected for this purpose are correlated with food
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23 470 production and derived from the Model of Ecosystem Dynamics, nutrient Utilisation,
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25 471 Sequestration and Acidification (MEDUSA 1.0) ecosystem model (Yool et al., 2011) and the
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27 472 Global Agro-ecological Zone (GAEZ) maximum potential yield of terrestrial crops under climate
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29 473 change (Fischer et al., 2012; Seo, 2014). The change from baseline to 2050 for each of the
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31 474 marine and terrestrial indicators can then be estimated. Since many coastal communities will also
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33 475 be engaged in other forms of economic activity such as agriculture a terrestrial exposure sub-
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35 476 component can also be estimated by comparing projected impacts of climate change on
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37 477 agricultural yields under different climate change scenarios. The LVI approach can then be used
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39 478 to collapse the marine and terrestrial sub-components into an exposure component. This can
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41 479 provide researchers with a modelled exposure component that can be examined in terms of the
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43 480 different climate change scenarios contained in the SRES/RCP.
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52 481 The added benefit of the integrated framework is that the vulnerability analyses output
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54 482 can be integrated with ecological vulnerability analyses being conducted for each hotspot by the
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3 483 other GULLS working groups. This allows for integrated, interdisciplinary analyses to be
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5 484 conducted for each hotspot. It also provides the potential to compare social vulnerability and
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7 485 adaptive capacity across and within hotspots. The framework's flexibility also allows for the
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9 486 comparison of quantitative models as well as qualitative data from other studies in the same
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11 487 hotspot study regions. Some of the outputs of the social vulnerability analyses can be used as
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13 488 inputs to ecological models, but can then be used to test the efficacy of alternative adaptation
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15 489 options through simulation testing. This is achieved using locally available and developed
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17 490 models for each region, including models of intermediate complexity (MICE) (Plagányi et al.,
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19 491 2014), Ecosim (Christensen and Walters, 2004) and Atlantis (Fulton, 2010). For example, simple
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21 492 MICE models are being developed for each region, with a focus on a few key fishery species and
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23 493 the communities that rely on these resources, in order to dynamically simulate and test coupled
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25 494 climate-biological-human interactions and responses. The MICE models are validated by fitting
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27 495 to available data over the historic period and then projected forward using climate projections
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29 496 from a high resolution global ocean model with biogeochemistry run under RCP8.5 scenario (the
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31 497 highest IPCC AR5 CO₂ emission scenario) to year 2099.

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38 498 A challenge for vulnerability analyses is ensuring an accurate in-depth assessment of
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40 499 vulnerability can be made at a scale that is relevant to decision makers. Although adaptation
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42 500 decision making occurs at many scales from national government down to individual households
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44 501 and this needs to be accounted for in vulnerability analyses (and in particular when designing
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46 502 adaptation options). Governance mapping has been undertaken in each hotspot which will
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48 503 subsequently be linked to the social ecological vulnerability data at different scales. Governance
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50 504 mapping is done through collection of formal laws, rules and regulations which map the top-
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52 505 down centralised maritime governance in each country. The social vulnerability survey and our
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3 506 engagement with marine-dependent communities provides the data to map the informal rules and
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5 507 governance of the marine resources and coastal areas in the study sites
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8 508 There are various analyses potentials and a key question in the analysis of data from
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10 509 vulnerability surveys is that of the appropriate weighting of indicators. The LVI framework
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12 510 provides a set of detailed guidelines for calculating a balanced weighted average composite
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14 511 index (Hahn et al., 2009; Mohan and Sinha, 2015). In the LVI approach, each indicator variable
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16 512 is standardized using the maximum and minimum values for the study population prior to being
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18 513 merged with all other indicators that contribute to a sub-component. The next step merges each
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20 514 of the sub-components that contribute to the relevant component using a consistent/standardized
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22 515 approach prior to using the components in the LVI equation. The approach used for the LVI
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24 516 means however that the estimates of vulnerability can only be compared to other estimates if
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26 517 they use the same method (Vincent, 2007). As highlighted earlier, the surveys conducted in each
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28 518 hotspot collected some consistent variables but others were adapted or removed from certain
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30 519 surveys. Furthermore, some surveys have had considerably larger numbers of respondents than
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32 520 others. Thus, the construction of vulnerability indices for each hotspot will have to be carefully
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34 521 considered to ensure that consistent and comparable results are derived. Indicators, sub-
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36 522 components and components may have to be weighted differently to the equal weighting used by
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38 523 Hahn et al. (2009) in order to estimate vulnerability in a consistent manner across hotspots.
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46 524 The addition of terrestrial exposure to climate change is an important component for this
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48 525 analysis for two reasons. First, many of the coastal communities currently relying on fisheries are
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50 526 likely to have a series of livelihood strategies in which the household engages. Thus, whilst
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52 527 fisheries are an important aspect of livelihoods in the hotspots the impact of climate change on
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54 528 terrestrial food production may also play an important part in the vulnerability of the
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3 529 communities. Thus, using the exposure components of land and marine will allow us to identify
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5 530 double hotspot regions (those that are projected to have high levels of exposure to negative
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8 531 marine and terrestrial change). Second, terrestrial agricultural production may provide an
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10 532 important adaptation option for some communities/hotspots. If this is not considered within the
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12 533 study in some way it is difficult to fully ascertain which communities/hotspots are likely to be
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14 534 the most vulnerable to climate-induced changes. The terrestrial projections of yield are based on
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16 535 different water supply options (rain-fed, irrigated) and different management options (High,
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18 536 intermediate and low intensity farming). Thus, providing a scenario system whereby the team
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21 537 can examine how a hotspot's vulnerability may change with a given development in agricultural
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23 538 input level or water supply system. The next stage will be to conduct group adaptation pathway
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25 539 development with each community to link with the outputs of the ecological and oceanographic
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28 540 GULLS models.
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33 34 542 **7. CONCLUSION**

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37 543 To develop a common methodology to compare social vulnerability across different
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39 544 communities, spanning multiple countries and ocean basins, an integrated yet flexible
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42 545 vulnerability framework has been developed. Not only it allows for robust comparisons of
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44 546 current and future vulnerabilities of coastal communities in different contexts to be made, but it
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46 547 avoids some of the typical shortcomings of social vulnerability research. It incorporates the
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48 548 social livelihood approach where other studies have only focused on the precepts of exposure,
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50 549 sensitivity and adaptive capacity which are not always easy to determine alone. Incorporating a
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52 550 flexibility component to the framework provides a greater sense of potential adaptability of
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55 551 individuals, occupations and institutions within the coastal and marine realm. The framework

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3 552 allows for future cross-scale comparisons, where instead of relying upon low resolution global
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5 553 datasets, data can be collected intensively at the local household level and can be analysed at that
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8 554 level but also integrated through indicators to allow for comparison to regional or national levels
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10 555 through Census data and even to the global level by integration with global indicators that map to
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12 556 the indicators from the surveys.

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15 557 Most importantly, the presented framework allows for seamless integration with the
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17 558 marine ecological system and the dynamics within. For instance, the value and relative
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19 559 importance of some of the indicators the ecological models produce (like fish abundance and
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21 560 biodiversity) can be established in the social vulnerability framework. It is possible to assess the
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23 561 consequences of a change in these crucial ecological indicators on the coastal communities or
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25 562 stakeholder groups. Finally, it addresses another major criticism of vulnerability analyses, which
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27 563 is the provision of a vulnerability score, but where no further work is carried out. The
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29 564 vulnerability analyses conducted by the GULLS hotspot teams will assist in providing the
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31 565 baseline and predictions of future vulnerability to develop sustainable and well-informed
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33 566 adaptation options with the study communities and countries.
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773 **Table 1** Comparison of vulnerability frameworks, their advantages, disadvantages, appropriate spatial scale and
 774 examples of their use in recent vulnerability studies.

Framework	Advantage	Disadvantage	Spatial scale	Focus area	Study
IPCC 2001*	Includes both ecological & social dimensions by using common concepts [exposure, sensitivity, adaptive capacity (AC)], Applicable to all spatial scales of vulnerability Comprehensive approach which is clear and understandable	Very broad level framework AC can be oversimplified into economic factors, disregarding social and ecological aspects	Global	National fisheries	Allison et al. 2009
			Regional	Indian Ocean coral reefs & communities	Cinner et al. 2012
			Local	Northern Gulf of California, Mexico	Morzaria-Luna et al. 2014
SLA/LVI	Multi-dimensioned view of poverty (economic and social) by using the 5 capitals Focused on aspects of sensitivity and AC Places high importance on policies and institutions	Provides no means to measure or integrate climate exposure or adaptation practices	Regional and Local	West African countries	Allison & Horemans 2006
			National	Australia	Metcalf et al. 2015
Food Security (FS)	Multi-dimensioned view of food-based welfare (can include: availability, stability, access and utilization) Easy access of required data	Most studies work from national level FAO datasets Food availability and access often focus of studies, neglect of other two dimensions Estimating food security at local level is difficult or not always relevant	Global	Global food security	Godfray et al. 2010
					Schmidhuber & Tubiello 2007
			Local	Household food security	Pinstrup-Andersen 2009
IPCC_LVI	Focuses on social dimension of vulnerability Index can be constructed from household data alone Doesn't depend on	Ecological vulnerability largely ignored	Regional	Mozambican regions	Hahn et al. 2009
				Ganges River basin	Mohan & Sinha 2015
			Local	Urchin fishery, California	Chen et al. 2014

	climate models to assess risk				
IPCC_LVI_FS	Strong social and ecological foundation	Large range of indicators required makes framework unsuitable for rapid vulnerability assessments	Global and local	Southern Hemisphere hotspots	Howard et al.
	Flexible framework allows comparison of communities across different levels of development				

*IPCC 2014 now utilises a different framework, where risk is central and vulnerability, hazards and exposure are three buttresses of risk that are then combined with climate and socio-economics (IPCC, 2014).

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For Review Only

800 **Table 2.** Vulnerability assessment survey method and logistical differences among hotspots

Country	Researchers	Survey format	Pre-testing	Field team no.	Survey language	Translators used	Training received	Sample size	Number of refusals/ unusable surveys	% of sampled households
Australia	3	Web*	Yes	0	English	No	N/A	104	21	0.12%*
Brazil	2	Field	Yes	1	Portuguese	No	N/A	151	23	65%**
India	3	Field	Yes	30	Malayalam	Yes	Yes	800	0	48%
South Africa	3	Field	Yes	1	Afrikaans	Yes	Yes	65	1	58%#
Solomon Islands	2	Field	No	2	Pijin English	Yes	Yes	110	0	30%
Madagascar	2	Field	No	2	Malagasy	Yes	Yes	46	0	55%**

801 * In southern Australia approximately 1 in every 3 households partake in recreational fishing activities each year (Lyle, Stark and Tracey 2014). There are
802 around 250,000 households which makes for a 0.12% sample.

803 ** The total number of marine dependent households per community is an estimation made by the researches before the survey application, as the official
804 number of marine dependent households in the area is not available. In Brazil the estimation is firstly based on the number of registered fishers per area and this
805 number was cross checked with the local fishing association and community leadership in terms of their estimate of the number of active fishers. Lastly it was
806 further verified through community site visits by the local researchers where visual estimates of the number of fishers were made.

807 # In South Africa and Madagascar the estimation was based on the number of fishing rights allocated in the sector (see Gammage et al 2017), which was
808 subsequently checked by the local researchers in field visits to the communities before the start of the survey.

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821 **Table 3:** Correlation coefficients greater than 0.5 for country level vulnerability and different country
 822 characteristics.
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Country characteristics	Variable name	Correlation coefficient	Interpretation
Size of the Exclusive Economic Zone	EEZ_size	-0.532	Smaller EEZ – greater vulnerability
Proportion of population who are undernourished	Undernourished	0.743	More people undernourished - greater vulnerability
Gross Domestic Product	GDP	-0.566	Lower GDP - greater vulnerability
Environment Protection Index	EPI_index	-0.632	Lower Environment Protection Index - greater vulnerability
Human Development Index	HDI_index	-0.575	Lower Human Development Index - greater vulnerability
Regulatory quality (score)	Regulatory_Quality	-0.599	Lower regulatory quality - greater vulnerability
Government effectiveness (score)	Govt_effectiveness	-0.823	Lower government effectiveness - greater vulnerability
Proportion of population for largest ethnic group	Majority_ethnicity	-0.744	Less ethnic diversity – greater vulnerability

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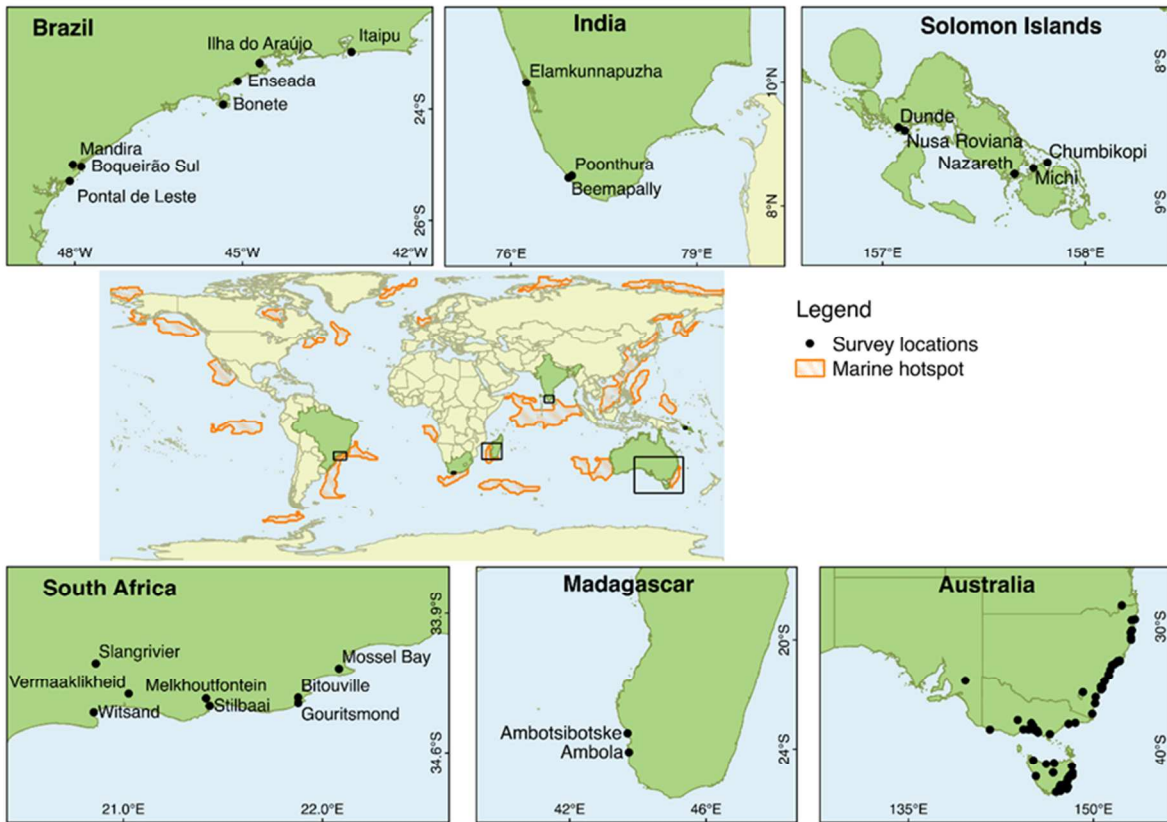
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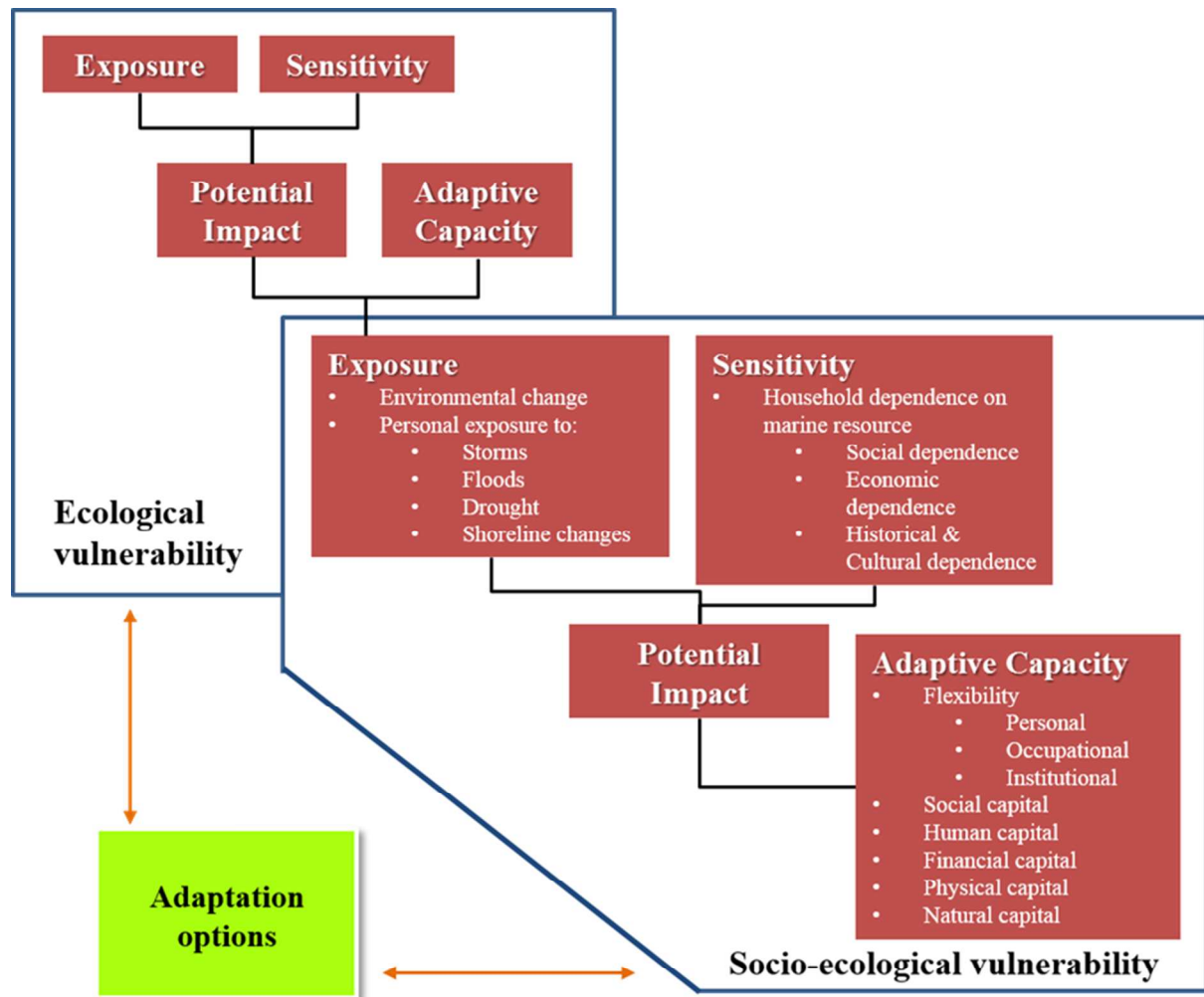
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 831 **Figure 1.** Countries where surveys were undertaken to assess human vulnerability to change in the marine
 832 environment. The survey locations are shown by black dots. In Australia the black dots represent the residence
 833 locations of individual respondents. The middle map shows the global marine hotspots.

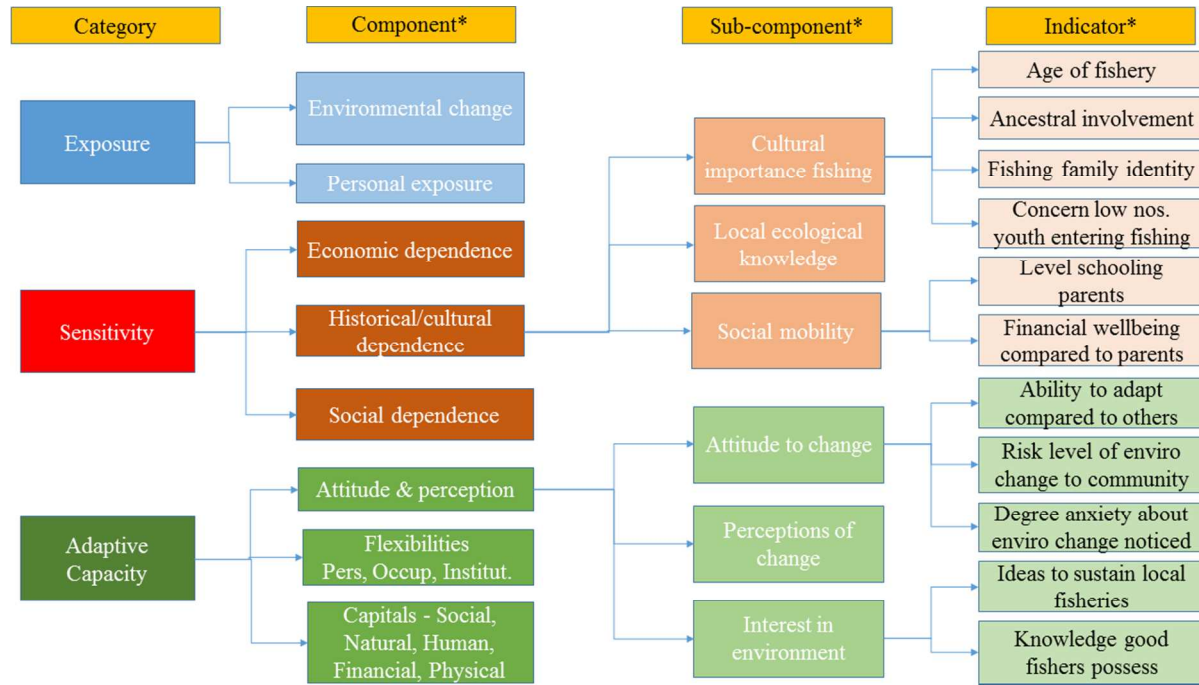
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836 **Figure 2.** GULLS framework for assessing coastal community vulnerability by integrating approaches from: IPCC,
 837 (2001); Chambers and Conway, (1992); Allison and Horemans, (2006); Marshall et al., (2013) and Metcalf et al.,
 838 (2015). Higher levels of complexity in the socio-economic subsystem are indicated showing different forms of
 839 dependencies, capitals and flexibilities.

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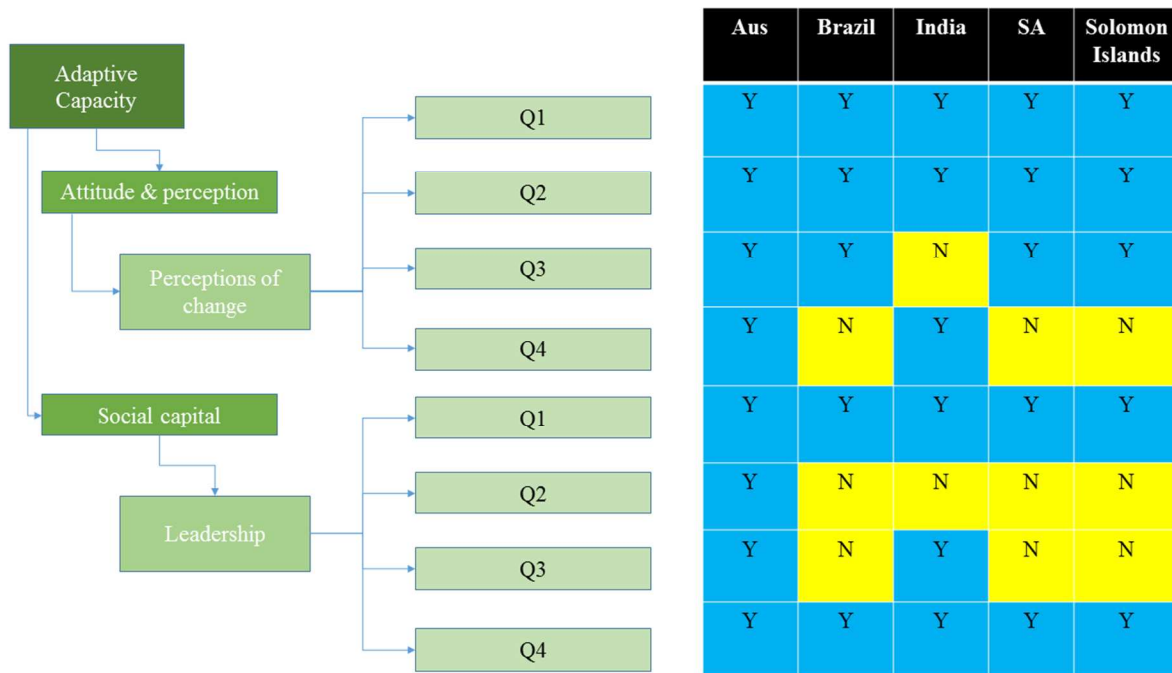


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 842 **Figure 3.** Structure of the integrated framework, demonstrating the multiple levels of complexity under each broad
 843 scale component and category.

844 *To ensure clarity of the figure, only a few components for each category are depicted and only some of the sub-
 845 components and indicators of each higher level are shown in this figure. Furthermore, due to space limitations,
 846 exposure is only shown up to the component level but in the full framework exposure does have components, sub-
 847 components and indicators.

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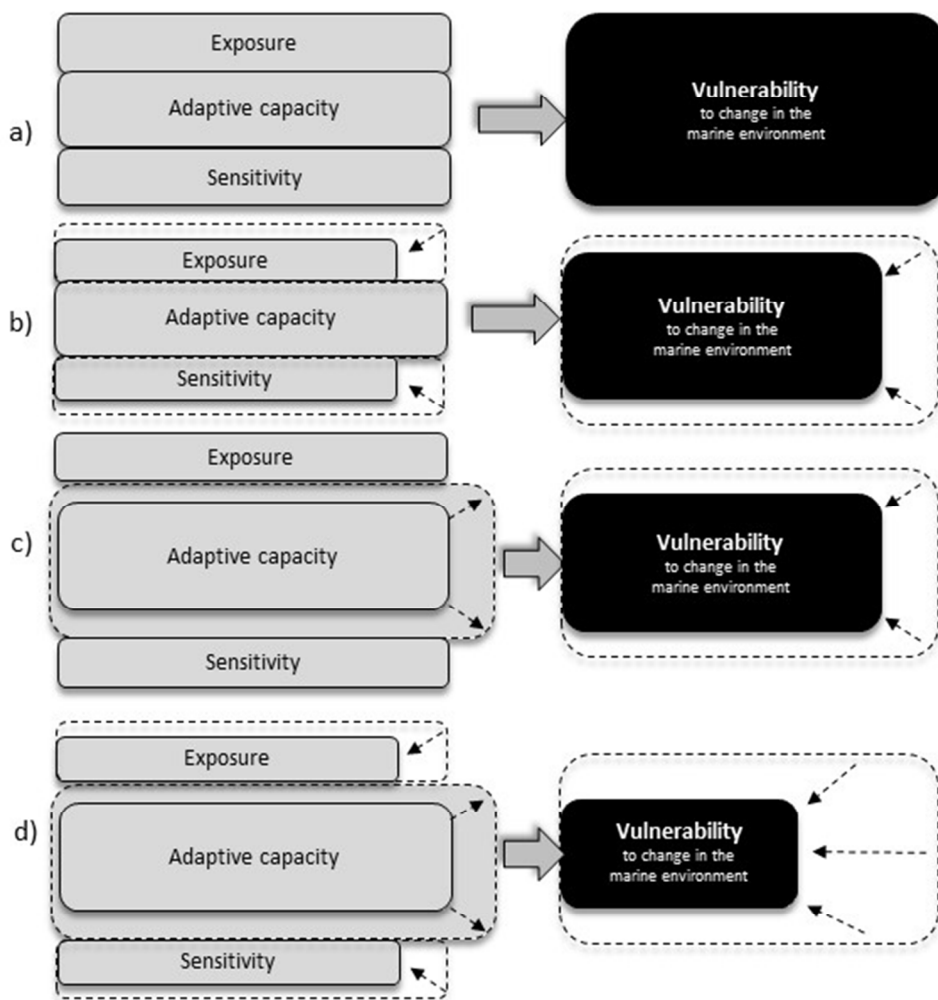
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852 **Figure 4.** The relationship between the conceptual vulnerability framework and the local survey instrument. Up to
 853 four different questions feed data back to each sub-component of the framework (LHS). Question selection (Y =
 854 Yes; N = No) differed across hotspot countries depending on cultural and contextual relevance (RHS).

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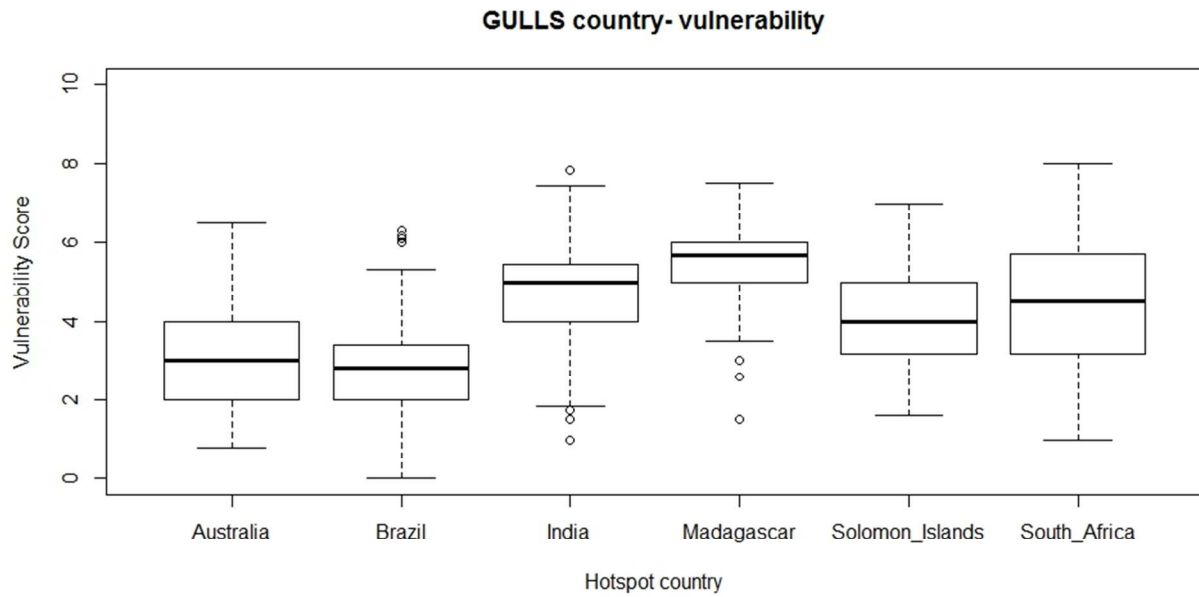


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858 **Figure 5.** Relationship between reducing vulnerability and reducing sensitivity and/or exposure and
 859 increasing adaptive capacity (adapted from Engle 2011)

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863 **Figure 6.** The vulnerability of hotspot countries to change in the marine environment. Scores range between 0 and
 864 10 where scores greater than 4.5 are worse. Higher scores indicate *greater* vulnerability.

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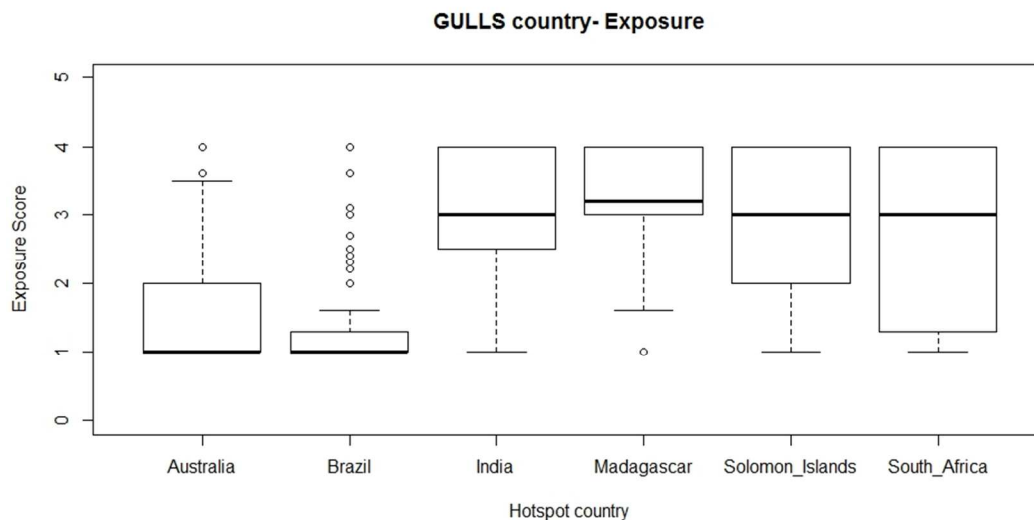
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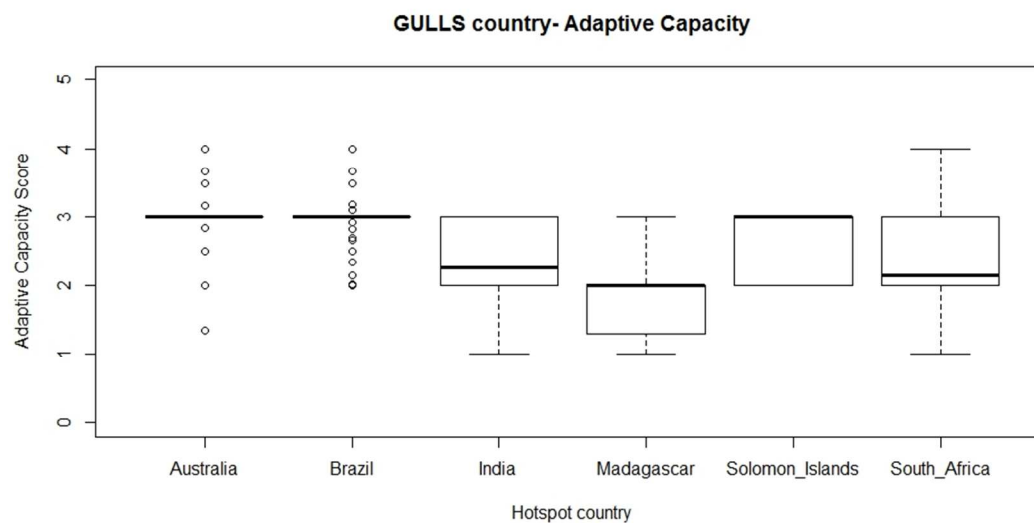
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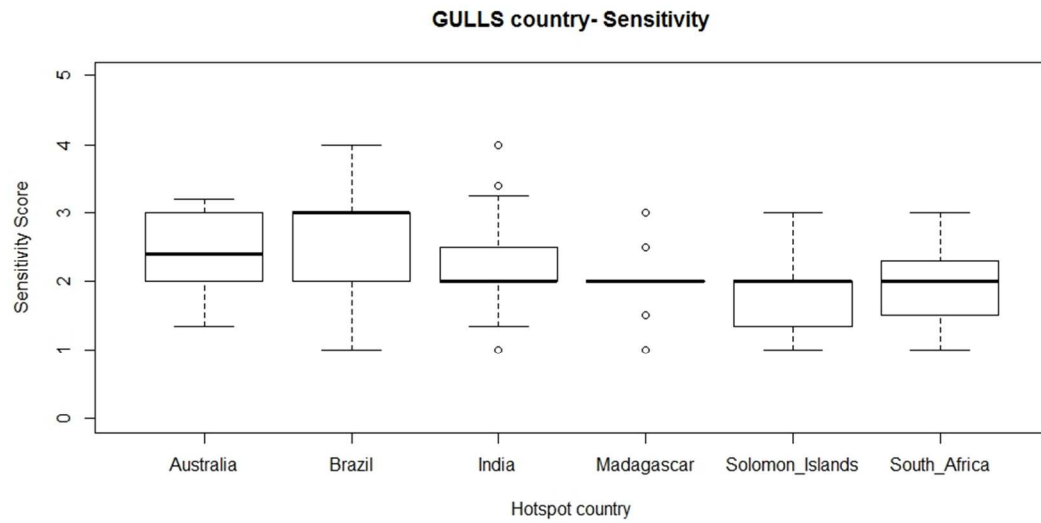
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874 b)



875 c)

876 **Figure 7.** The a) exposure b) adaptive capacity, and c) sensitivity of hotspot countries to change in the marine
 877 environment. Scores range between 0 and 5 where scores greater than 2.5 are worse. Higher scores indicate *greater*
 878 exposure and sensitivity (which is not beneficial), and higher scores also indicate greater adaptive capacity (which is
 879 beneficial).

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