
Fishing for euros: how mapping applications can assist in maintaining revenues under the Landing Obligation

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Abstract :

With the introduction of the Landing Obligation (LO) in EU fisheries, there is an increasing need for fishers to avoid unwanted catches while maximizing revenues. Improving understanding of the spatio-temporal dynamics of unwanted catches could assist the fishing industry optimize catches by altering where they fish. How following such advice relates to revenues and fishery dynamics requires more consideration. We take an existing hotspot mapping methodology and examine how it could be used to identify fishing opportunities under the LO in Irish (Celtic Sea) and Danish (North Sea and Skagerrak) demersal fisheries. We consider if fishing effort can be relocated to avoid unwanted catches while maintaining revenues. The value per unit effort of fishing activity in both areas was often linked to high catch rates of key demersal species (cod, haddock, hake, and whiting). Our analyses indicated, however, that there are options to fish in areas that could provide higher revenues while avoiding below minimum conservation reference size catches and choke species. This was evident across both case study areas demonstrating that hotspot mapping tools could have wide applicability. There does, however, remain a need to explore how the displacement of vessels may further alter species distributions and fleet economics.

Keywords : discards, economics, fishing tactics, hotspot maps, Landing Obligation, spatial avoidance

26 INTRODUCTION

27 Discards refer to catches that are returned to the sea during fishing operations. They have long been
28 acknowledged as a widespread part of global fishing operations (Alverson *et al.*, 1994; Jennings *et al.*,
29 2001; Borges *et al.*, 2005; Catchpole *et al.*, 2014) with 9.1 million tonnes of discards being reported
30 from marine capture fisheries between 2010 and 2014 (Pérez-Roda *et al.*, 2019). The issue of
31 discarding has become a global concern in recent years with the incidental removal of resources
32 threatening the sustainability of many species in addition to representing a waste of a rich source of
33 dietary protein (Kelleher, 2005; Catchpole and Gray, 2010; Bellido *et al.*, 2011; Little *et al.*, 2015;
34 Catchpole *et al.*, 2017). The gradual elimination of discards and unwanted catch in European fisheries
35 was, therefore, identified as one of the main objectives of the EU Common Fisheries Policy reforms in
36 2013. This resulted in the gradual introduction of the Landing Obligation (LO) from 2015,
37 with this legislation being fully implemented since the start of 2019. The LO which has prohibited
38 the prohibits the discarding of species subject to TACs (total allowable catches) and size limits, with
39 some exemptions for species which have high survivability in addition to de minimis allowances,
40 which allow for a discard fraction of up to 5% in fisheries when increased selectivity is difficult to
41 achieve ~~since the beginning of 2019~~ (European Commission, 2013; Catchpole *et al.*, 2014).

42 There are numerous reasons why fishers discard catches, from management constraints to
43 environmental, economic and social drivers (Alverson *et al.*, 1994; Hatcher, 2014; Rochet *et al.*, 2014;
44 Milisenda *et al.*, 2017). From an economic perspective it is often suggested that fishers are profit
45 maximizers and will discard any element of their catch that is not profitable to land (Van Putten *et al.*,
46 2012; Batsleer *et al.*, 2015; Hatcher and Drakeford, 2015). It is also assumed that fishers will target
47 areas of best catch, in terms of greatest catch rate and profitability (e.g. Holland and Sutinen, 1999;
48 Wilen *et al.*, 2002; Tsitsika and Maravelias, 2008). With the introduction of the LO there is an increased
49 incentive to avoid species that could 'choke' a fishery (Schrope, 2010). The expectation is that fishers
50 operating in mixed fisheries in particular, where vessels operating a variety of gears target several

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3 51 different species, will modify their fishing behaviour (through technical and/or spatio-temporal
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5 52 changes) in order to avoid unwanted catches and comply with the prohibition. Whilst advances in gear
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7 53 technology provide methods to increase selectivity it is almost impossible to fully eliminate all
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9 54 unwanted catch from commercial gears in mixed fisheries while maintaining profitable fishing activity
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11 55 (Guillen *et al.*, 2018; Pointin *et al.*, 2019; Reid *et al.*, 2019; Rodríguez-Rodríguez *et al.*, 2019; Suuronen
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13 56 and Gilman, 2019). In addition to technological developments, therefore, it is assumed that the LO,
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15 57 and the need to avoid unwanted catches, may alter the spatio-temporal patterns of commercial
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17 58 fishing operations.

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22 59 The importance of better understanding the spatio-temporal distribution of bycatch and adopting
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24 60 spatial avoidance techniques have been recognised by industry and scientists alike (Dunn *et al.*, 2011;
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26 61 Paradinas *et al.*, 2016). In recent years numerous methods have utilised fisheries and vessel location
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28 62 data to identify and predict how catch compositions are likely to vary over space and time and how
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30 63 this relates to vessel dynamics (Fraser *et al.*, 2008; Lewison *et al.*, 2009; Lee *et al.*, 2010; Paradinas *et*
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32 64 *al.*, 2016; Bellido *et al.*, 2019; Calderwood *et al.*, 2019; Reid *et al.*, 2019). Scientists are now developing
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34 65 web-based applications to display the data produced by such methods in a digestible format, making
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36 66 the data more accessible to industry stakeholders (Calderwood *et al.*, 2019; Reid *et al.*, 2019). As these
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38 67 tools are developed it is important to understand how the information within them relates to the
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40 68 many economic drivers of fishing behaviour. Certainly with the introduction of the LO there is a need
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42 69 to determine how any relocation of fishing effort to avoid unwanted catches may impact upon the
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44 70 profitability of fisheries, -with regard to both fisheries that relocate activity and those that occur in
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46 71 areas that experience an influx of new fishing activity (Pointin *et al.*, 2019). Additionally, a number of
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48 72 these tools have been developed on a case-by-case basis, concentrating on specific gear types or
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50 73 fisheries in specific geographic areas, using different methodologies for the spatial analyses.
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52 74 Furthering our understanding of how these methodologies and tools are applicable across multiple
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54 75 fisheries is also important as such work develops in the future (Paradinas *et al.*, 2016; Pennino *et al.*,
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56 76 2017).

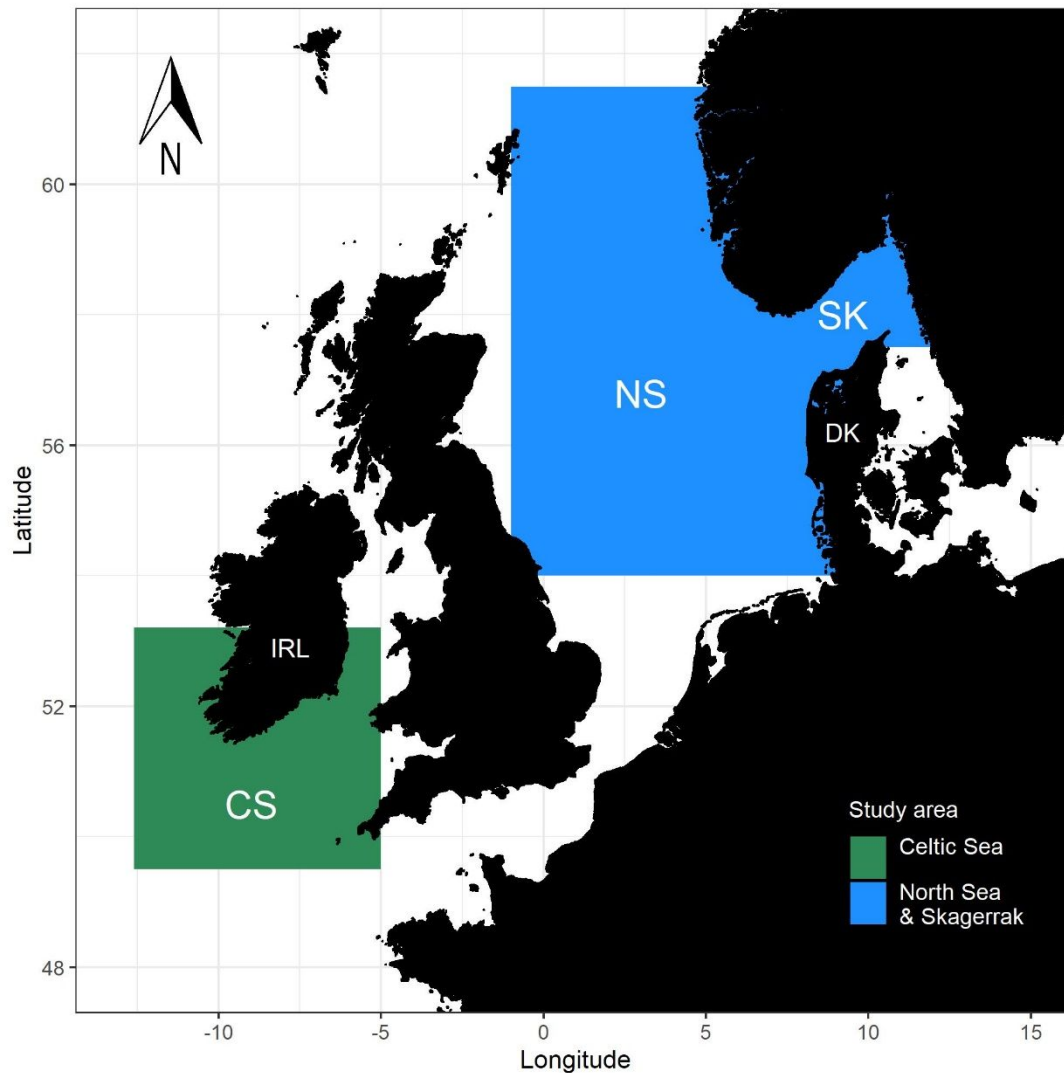
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3 77 In this paper we aim to determine if it may be possible to relocate fishing effort to avoid unwanted
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5 78 catches while maintaining revenues within a fishery. This was achieved by first determining whether
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7 79 a link existed between the economics of fishing and discarding behaviour prior to the introduction of
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9 80 the LO. We then examine how the mapping of potential discard hotspots, designed to assist fishers in
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11 81 avoiding unwanted catches, also relate to vessel and fleet economics. The potential economic losses
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13 82 or gains associated with avoidance behaviours are examined. The potential displacement of fishing
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15 83 effort, if information in these maps were to be adhered to, is also considered. This work focuses on
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17 84 two case study fisheries: Irish demersal trawlers operating in the Celtic Sea and Danish demersal
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19 85 trawlers operating in the North Sea and Skagerrak.
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27 87 **METHODS**

28 29 30 88 **Case study fisheries**

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33 89 Data collected from Irish and Danish demersal trawlers operating in the Celtic Sea and the North Sea
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35 90 and Skagerrak were compared (Figure 1). In both cases only data from vessels operating TR1 gears,
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37 91 which are defined as bottom trawls, Danish seines and similar towed gears, excluding beam trawls,
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39 92 with cod end mesh sizes above 100mm (European Commission, 2008; Davie and Lordan, 2011), were
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41 93 used. In the Irish fleet the majority of TR1 vessels operate otter trawls and target a mixed demersal
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43 94 whitefish fishery. The Irish fleet is subject to individual monthly quotas with no opportunities for quota
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45 95 swapping or sharing (DAFM, 2016a; Calderwood and Reid, 2019). The Danish TR1 vessels mainly
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47 96 operate otter trawls like the Irish, but in addition to whitefish, species like European plaice, anglerfish
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49 97 and *Nephrops* are also valued and targeted catches (Mortensen *et al.*, 2017, 2018; Plet-Hansen *et al.*,
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51 98 2018). The Danish fleet is governed by an Individual Transferable Quota (ITQ) system, in which some
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53 99 fishers have subsequently formed quota pools to lease quota from other vessels or from a common
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58 100 quota pool (Andersen *et al.*, 2010; Mortensen *et al.*, 2018). Due to the low coverage of observer data,
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3 101 which feeds into the mapping methodology used, selecting data based on the TR1 definition provides
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5 102 sufficient data to formulate species hotspots and allow for comparison of vessels ~~targetting~~targeting
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7 103 similar demersal fisheries in the Celtic Sea and North Sea and Skagerrak.
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105 Figure 1: Map of Northwestern Europe with the study areas. Green colour indicates the Celtic Sea (CS), blue
106 colour indicates the North Sea (NS) and Skagerrak (SK). IRL = Ireland, DK = Denmark.
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107 Irish and Danish data

108 Discard hotspots and VPUE data

109 Mean value per unit effort data (VPUE ~~-- value of landed catch (€)euros earned~~ per hour of fishing
110 activity for total landed catch of all species caught) were calculated using VMS (vessel monitoring

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3 111 system) and logbook data for all TR1 vessels operating in the two case study areas. Logbook data
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5 112 provided for both fisheries contained information on total weight of catches for TAC species, ~~and~~
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7 113 ~~the~~ monetary value of these catches were provided by the SFPA (Sea Fisheries Protection Agency)
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9 114 alongside logbook data for the Irish fisheries. In the Danish fishery the Danish Fisheries Agency provide
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11 115 the same economic data from sales slips alongside logbook records. The VMStools package in R
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13 116 (Hintzen *et al.*, 2016) was used to remove erroneous data from the VMS dataset and to merge VMS
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15 117 and logbook data (Hintzen *et al.*, 2012; R Development Core Team, 2012). VMStools was then used to
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17 118 allocate Ddaily catch weight and sales value data equally ~~were equally allocated~~ to each location along
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19 119 a vessel's recorded path where the associated vessel was recognised as being engaged in fishing
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21 120 operations, based on vessel's speed (Gerritsen and Lordan, 2011; Hintzen *et al.*, 2012). Value data
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23 121 were standardized to the value of catches per hour of fishing activity, with both catch volume and
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25 122 value being equally allocated across all fishing activity per trip. The resultant data were then allocated
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27 123 to 0.2° x 0.2° rectangles and mean values calculated for each cell. Mean values were first calculated
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29 124 using data collected between 2010 and 2015 to allow for comparison with discarding hotspots prior
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31 125 to the introduction of the LO. This process was repeated to produce mean values for data collected
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33 126 from 2016 to be used when assessing the usefulness of hotspot maps as a discard avoidance tool.
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35 127 Data collected by onboard observers as part of the EU data collection framework (Council regulation
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37 128 (EC) No 199/2008) were used to determine discarding hotspots within both the Irish and Danish
38
39 129 demersal fisheries for four key TAC species: cod (*Gadus morhua*), haddock (*Melanogrammus*
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41 130 *aeglefinus*), hake (*Merluccius merluccius*) and whiting (*Merlangius merlangus*). These four species
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43 131 were chosen based on their importance in both case study areas in terms of a combination market
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45 132 value and abundance. Please see [Supplementary Material 1](#), ~~—~~for more detail. On-board sampling
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47 133 protocols followed those described by Håkansson (2019) and Borges *et al.* (2005a), thus allowing for
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49 134 data on the weight of TAC species discarded to be raised to haul level using length weight keys for
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51 135 each species sampled. In the Celtic Sea Ddata were collected from a total of 2674 individual hauls from
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53 136 226 trips, with an average trip length of 6.1 days, which took place on 50 different vessels between
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3 137 ~~2010 and 2015. the Celtic Sea~~In the North Sea and Skagerrak data were collected from and 352 hauls
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5 138 from ~~250 trips, with an average trip length of 5.1 days, from 127 unique vessels the North Sea and~~
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7 139 ~~Skagerrak~~ between 2010 and 2015. Danish fisheries observer data were supplemented by Electronic
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9 140 Monitoring (EM) data for 4 vessels, providing information from a total of 8 individual hauls from 8
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11 141 trips, with an average trip length 7.6 days. Although this is a low number, these 4 vessels would
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14 142 otherwise not have any records as they had not conducted trips with an observer in the study period.
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16 143 EM sampling followed the descriptions by Bergsson *et al.*, 2017 and Plet-Hansen *et al.*, 2019, collecting
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18 144 length and weight estimates at the haul level.
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23 146 Using the hotspot mapping methodology described by Calderwood *et al.*, (2019) the 2010-2015
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25 147 observer data were used to highlight discarding patterns prior to the introduction of the LO for any
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27 148 demersal species. First the discarding rate (kg hr^{-1} of discards) was calculated per haul for each of the
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29 149 four TAC species selected. All data were then assigned to $0.2 \times 0.2^\circ$ grid cells, based on the
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31 150 geographical midpoint of each haul. Mean annual discard rates were calculated per grid cell and
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33 151 subsequently binned into five equal quantiles, following the removal of zero discard values, for which
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35 152 a separate category was assigned. Finally, an amalgamated map for 2010-2015 was created for each
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37 153 species by identifying grid cells that were consistently within the same binned category over multiple
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39 154 years. Thus, areas that displayed variability over multiple years were removed, leaving just those areas
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41 155 that showed consistent discarding rates over time (Calderwood *et al.*, 2019). Grid cells consistently
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43 156 containing the highest category of discard rates (top 40% of values) were selected and overlaid onto
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45 157 the gridded VPUE data for the period 2010-2015. To determine whether there was a relationship
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47 158 between the occurrence of high discarding hotspots and the value of total catches, the mean VPUE
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49 159 data were also binned into five equal quantiles. A chi-squared test was performed to compare the
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51 160 number of occurrences of discarding hotspots within each of these five VPUE categories for each
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53 161 species individually, and also for all four species together.
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163 **Total catch hotspots and VPUE data**

164 Gridded hotspot maps were also created for the total catch rate (kg hr^{-1} caught) for both the above
165 (>) and below (<) MCRS (minimum conservation reference size) components for the four key TAC
166 species. The purpose of these maps is to use observer data to identify consistencies in catch patterns
167 over multiple years, providing a tool to fishers to better inform decision making on where to fish to
168 optimise opportunities under the LO. In the Celtic Sea a tri-national observer data set (French, British
169 and Irish data) was used to create these maps as described in Calderwood *et al.*, (2019). The same
170 approach was repeated for the North Sea and Skagerrak using EM data (Plet-Hansen *et al.*, 2019) and
171 observer data from Danish vessels only.

172
173 To test how information contained within these maps might relate to future fleet economics, if such
174 maps are to be used as a predictive tool to determine optimum fishing locations under the LO, we
175 compared the two metrics. The mean VPUE of all species landed per 0.2° by 0.2° grid cell for 2016 was
176 compared with the binned hotspot mapping category of the same grid cell calculated using data from
177 2010-2015 for each species and size class of interest (< and >MCRS cod, haddock, hake and whiting).

178 Data from 2010-2015 were used to create the hotspot maps to take account of all catches prior to the
179 introduction of any LO legislation with regard to demersal fisheries to build up a picture of consistent
180 catch patterns prior to the introduction of this legislation. In 2016 some LO restrictions were applicable
181 to TR1 vessels operating in both case study areas. In the Celtic Sea in 2016 vessels were required to
182 land all whiting if in 2013/2014 >25% of their landed catch had consisted of cod, haddock, whiting and
183 saithe combined, or land all hake if >30% of their landed catch in 2013/2014 had consisted of this
184 species (DAFM, 2016a). In the North Sea and Skagerrak TR1 vessels were required to land all haddock
185 and plaice. TR1 vessels with >50% of all landings comprised of saithe during 2012-2014 would be
186 categorised as a saithe targeting vessel and would also be required to land all catches of saithe as well
187 (European Commission, 2015a). In the Danish EM trial in the North Sea and Skagerrak, no significant

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3 188 [change in discards of haddock was registered in 2016 compared to 2015](#) (Plet-Hansen *et al.*, 2019).
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5 189 [These introductions, may have had some impact on fishing behaviour, as recorded by VMS and](#)
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7 190 [logbook data in 2016. A comparison of the hotspot maps, created with observer data from 2010-2015,](#)
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9 191 [with fishing data from the following year, however, still proves a valuable insight into how hotspot](#)
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11 192 [maps relate to the spatial variation in subsequent catch values and how this information could](#)
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13 193 [potentially be utilised following the full implementation of the LO. To achieve this](#) ~~M~~mean VPUE data
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16 194 were tested for normality and homogeneity of variance prior to analysis using a number of tests as
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18 195 well as through the visual inspection of residuals. Normality was tested using Shapiro Wilk's test with
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20 196 the homogeneity of variance then being tested using either a Bartlett or Levene's test depending on
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22 197 whether the data were deemed to be normal or not (Underwood, 1997). In all cases data did not
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24 198 conform to the assumptions required for linear regression analysis so general linear modelling (GLM)
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26 199 was run, fitted with a gamma distribution and a log link function to account for the positive skew of
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28 200 the data (Zuur *et al.*, 2009). Hotspot mapping categories were defined as ordered factors with the
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30 201 species and size category of the corresponding map fitted as additional fixed factors in the GLM. Post-
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32 202 hoc Tukey tests were used to make comparisons among levels of significant terms.
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204 **Below MCRS avoidance scenarios**

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44 205 Under the LO all catches of TAC species will have to be landed regardless of size, but only those catches
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46 206 >MCRS can be sold for human consumption (European Commission, 2015b) . To maximise revenue, it
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48 207 would be important to target >MCRS catches while avoiding <MCRS fish (Table 1). We therefore
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50 208 assessed how the hotspot maps could be used to avoid <MCRS catches of the four key TAC species
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52 209 while targeting >MCRS catches of the same species. We then determined how this spatial avoidance
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54 210 could impact VPUE. The areas identified as consistently being in the top two categories in CPUE
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56 211 hotspot maps (top 40% of catches by weight) for <MCRS and >MCRS catches respectively were
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58 212 overlaid for each species. Grid cells were then identified either as being in the <MCRS or >MCRS
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3 213 category or as an area where high catch rates of these two size classes overlapped. Total vessel activity
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5 214 and associated VPUE were extracted from each relevant grid cell using 2016 logbook and VMS data.
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7 215 The mean 2016 VPUE data associated with each cell were again tested for normality and homogeneity
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9 216 of variance as previously described but they did not meet the assumptions of analysis of variance
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11 217 (ANOVA) even following data transformation. Due to the positive skew of the data, they were again
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13 218 analysed with a GLM fitted with a gamma distribution and log link function (Zuur *et al.*, 2009). The size
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15 219 category (three levels: <MCRS, >MCRS, overlap) was fitted as a fixed factor in the GLM. To account for
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17 220 the small and uneven sample sizes associated with each category the analysis was run using Type III
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19 221 sum of squares (Quinn and Keough, 2002).
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26 223 **Species avoidance scenarios**

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28 224 The above procedure was applied to different cases illustrating one common choke species in the Irish
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30 225 fishery and one common choke in the Danish fishery (Schrope, 2010). For the Irish fishery the analysis
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32 226 focused on the overlap of the top two CPUE categories for the >MCRS components of whiting, which
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34 227 currently has reasonably large quota available, and haddock, which often has limited quotas available
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36 228 (Calderwood *et al.*, 2016). For the Danish fishery we focused on the optimization of >MCRS cod catches
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38 229 (an important target species for Danish TR1 vessels in the North Sea) and the avoidance of hake, which
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40 230 is currently increasing in the North Sea. This has resulted in the TAC being relatively low compared to
41
42 231 the spawning stock biomass ~~both~~ because of the time lag in stock assessments, the relatively fast
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44 232 northward shift of the distribution in hake in this areas (Kraak *et al.*, 2013; Baudron and Fernandes,
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46 233 2015) and because historical catches of hake were low in the North Sea when catch shares were
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48 234 allocated within EU member states (Baudron and Fernandes, 2015). Additionally, the price per kg of
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50 235 hake is generally lower, in the Danish markets, than for cod ~~and haddock for instance~~, and the fish
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52 236 also has a tendency to be damaged during the haul, due to abrasion and pressure from the
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54 237 surrounding catch, which further reduces its sale value (Catchpole *et al.*, 2018; Plet-Hansen *et al.*,
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56 238 2019). Thus the top two CPUE categories for the >MCRS components of cod and hake were plotted
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239 and analysed. All analyses were undertaken in R 3.2.5 (R Core Team, 2017) and an overview of the
240 data used in each section of the analyses is presented in Figure 2.

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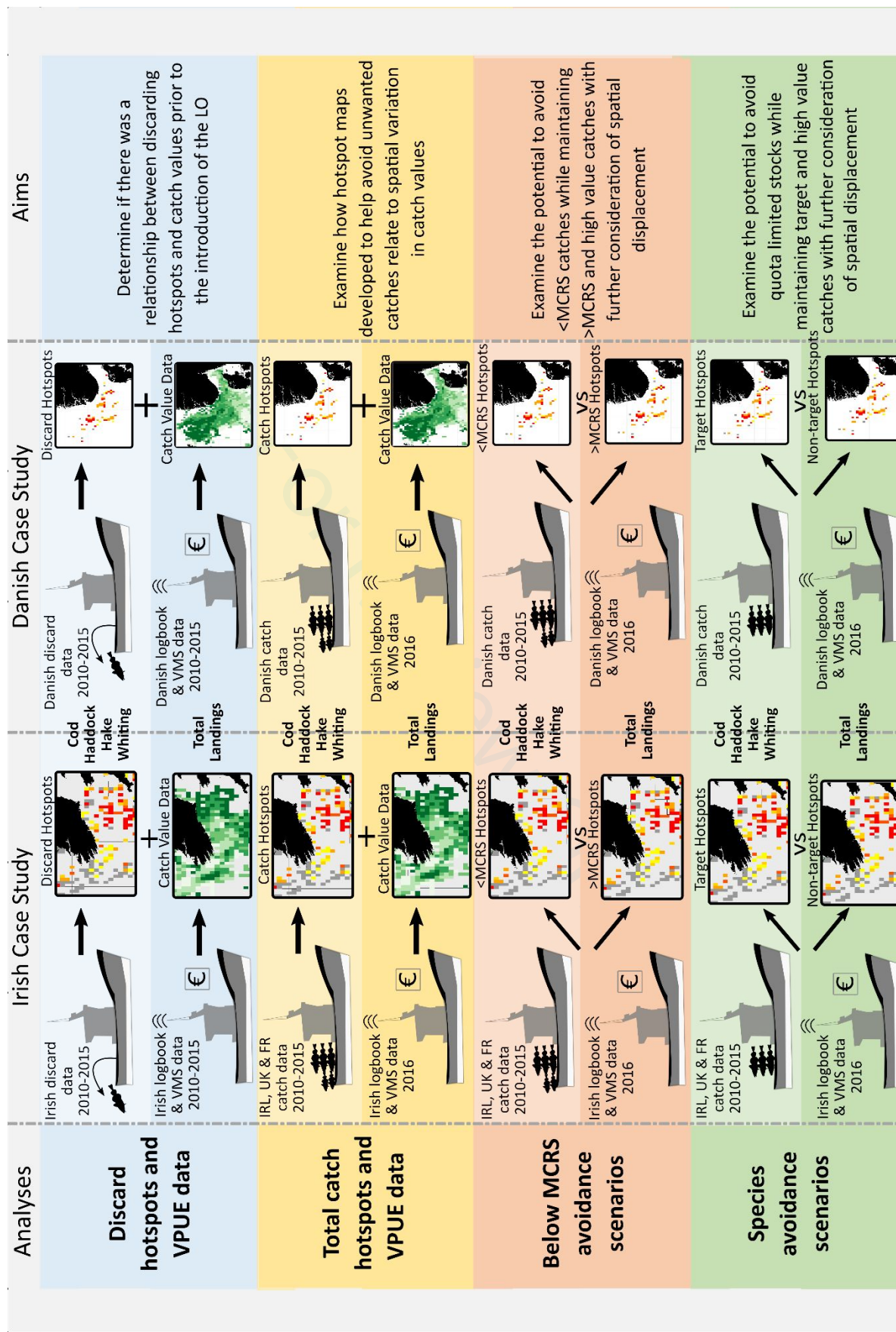


Figure 2: A schematic diagram to show the data inputs for each of the four sections of the analysis.

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Table 1: The Minimum Conservation Reference Size (MCRS) in cm for the key TAC species considered in this paper.

Species	Minimum Conservation Reference Size (cm), Celtic and North Sea	Minimum Conservation Reference Size (cm), Skagerrak
Cod	35	30
Haddock	30	27
Hake	27	30
Whiting	27	23

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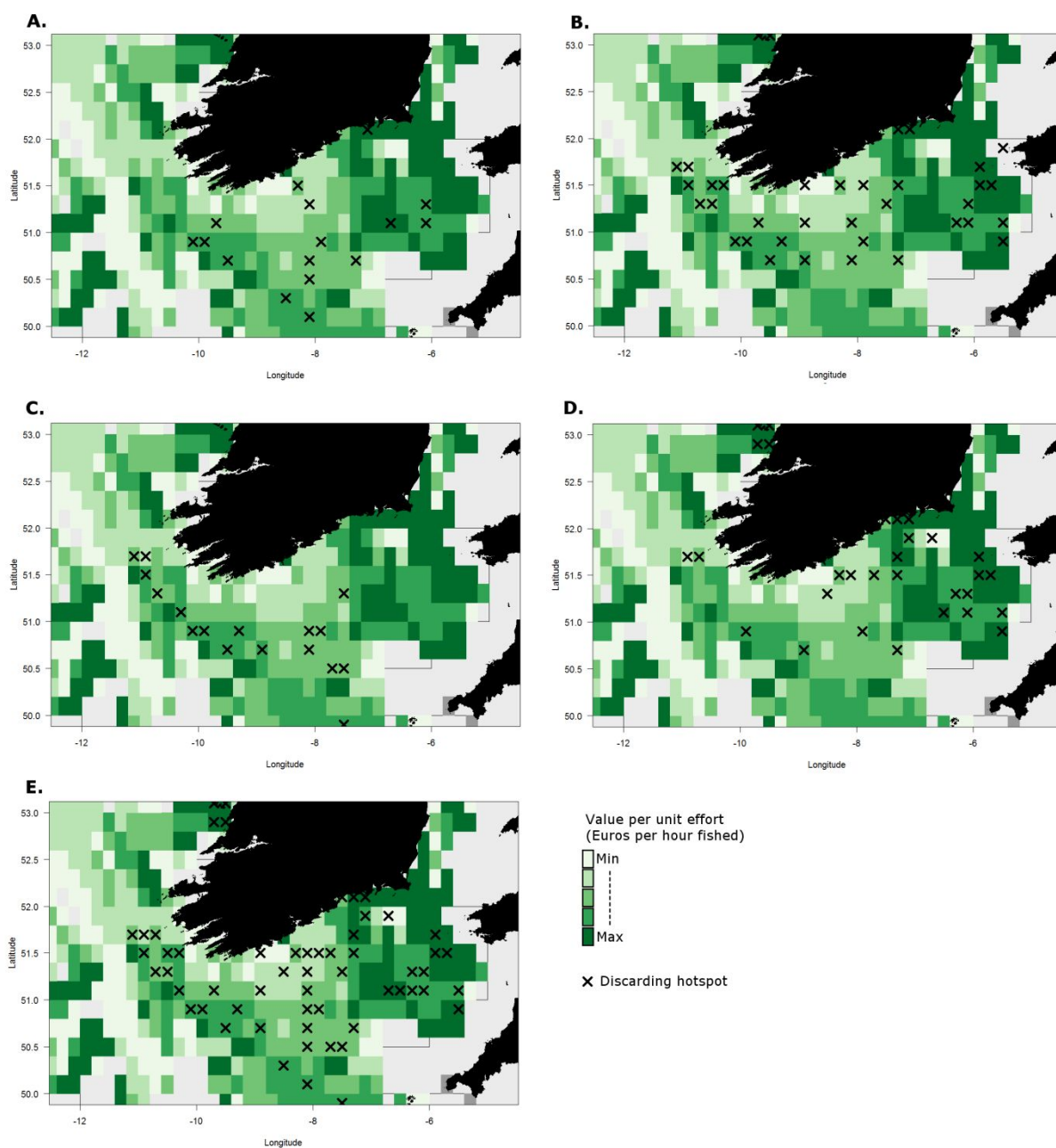
250 RESULTS

251 Discarding Hotspots and VPUE

252 There are spatial differences in where peak quantities of discards occur for the four key commercial
 253 species highlighted for both Irish and Danish fisheries. There are also differences in the relationship
 254 between discarding hotspots and landing values. In the Celtic Sea, discarding hotspots for both
 255 haddock and whiting are numerous and spread across the area. The majority of hotspots for both
 256 species have been identified between 5° and 11°W and 50.5° and 52.5°N although hotspots for whiting
 257 are concentrated to the east of 9°W (Figure [2B-3B](#) and [2D3D](#)). The majority of cod in the Irish fishery
 258 is consistently discarded in the southeast region of the study area with most discarding hotspots being
 259 below 51.5°N and/or east of 10°W (Figure [2A3A](#)). Hake discards were also more concentrated south
 260 of 51°N (Figure [32C](#)). There was a significant relationship between the occurrence of discarding
 261 hotspots and the mean catch value for all species in the Irish fishery (Supplementary [1A2A](#); Table 2).
 262 In the case of whiting the greatest number of discarding hotspots were associated with the highest

263 value fishing areas. Yet only 13 discarding hotspots were associated with these high value areas,
 264 representing- 11.4% of such areas. The occurrence of discarding hotspots for cod, haddock, hake and
 265 the four species combined were mainly associated with the second highest value category,
 266 representing 8.8%, 11.4%, 9.6% and 26.3% of the total area associated with that value category,
 267 respectively.

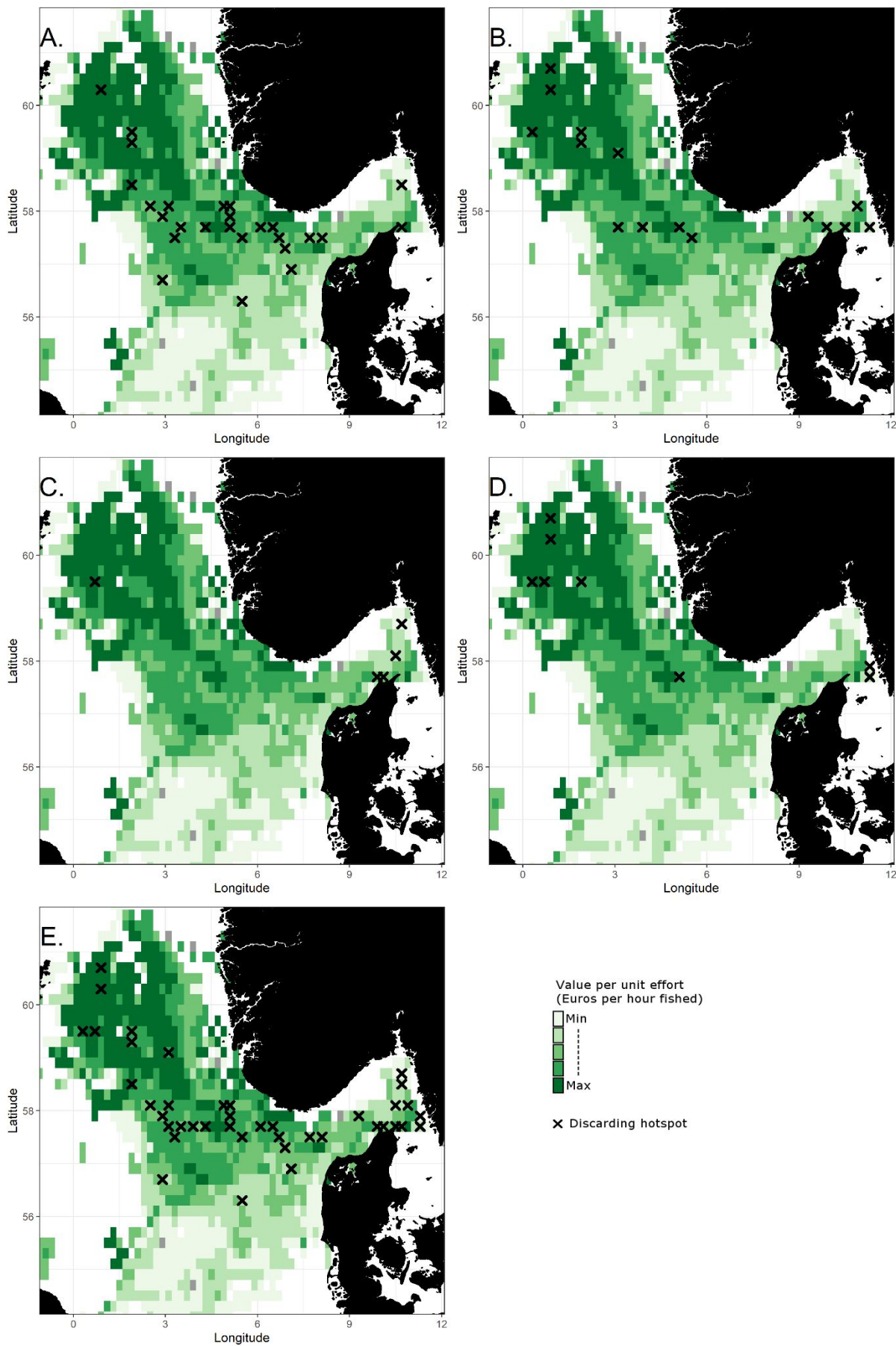
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3 270 Figure 32: Maps showing the mean value of total catches (euro per hour) for the Irish fleet in the period 2010 -
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5 271 2015 related to discarding hotspots identified during the same period for A. cod, B. haddock, C. hake, D. whiting
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7 272 and E. cod, haddock, hake and whiting combined.
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3 274 Figure 43: Maps showing the mean value of total catches (euro per hour) for the Danish fleet in the period 2010-
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5 275 2015 related to discarding hotspots identified during the same period for A. cod, B. haddock, C. hake, D. whiting,
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7 276 E. cod, haddock, hake, and whiting combined.
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13 278 For the Danish fishery, cod discarding hotspots tend to be closer to the south and southwest coast of
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15 279 Norway as well as in the central North Sea. Haddock discarding hotspots overlap with several cod
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17 280 discarding hotspots in the northern North Sea but also occur along the 58th parallel north (Figure 43A
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19 281 and 3B). Except for one discarding hotspot, all hake hotspots occur in Skagerrak (Figure 43C). Whiting
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21 282 discarding hotspots are either at the boundary between Skagerrak and Kattegat or overlapping with
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23 283 haddock discarding hotspots, mainly in the northern North Sea (Figure 43D). An area between the
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25 284 Shetland Islands and Norway is low in discard hotspots for the Danish case, even when plotting discard
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27 285 hotspots for all four study species in one map (Figure 43E). In the Danish case a significant relationship
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29 286 was found between VPUE and discarding hotspots for cod, haddock and the combined species map
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31 287 only (Supplementary 1B2B; Table 2). Both cod and haddock had the highest number of discarding
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33 288 hotspots associated with grid cells in the second highest value category, the hotspots only overlapped
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35 289 with areas identified with the highest VPUE in 1.4%, 1.4% and 3.3% of cases respectively. It can be
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37 290 seen on the maps that this is mainly because of a large area in the northern North Sea with high value
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39 291 cells and no discarding hotspots identified (Figure 43).
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56 296 Table 2: χ^2 results testing the relationship between the value of catches for the Irish fleet operating in the Celtic
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58 297 Sea and the Danish fleet operating in the North Sea and Skagerrak in 2010 and 2015 and the association of
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discarding hotspots for cod, haddock, hake, whiting and a combination of these species. Significant value ($p < 0.05$) are indicated in bold.

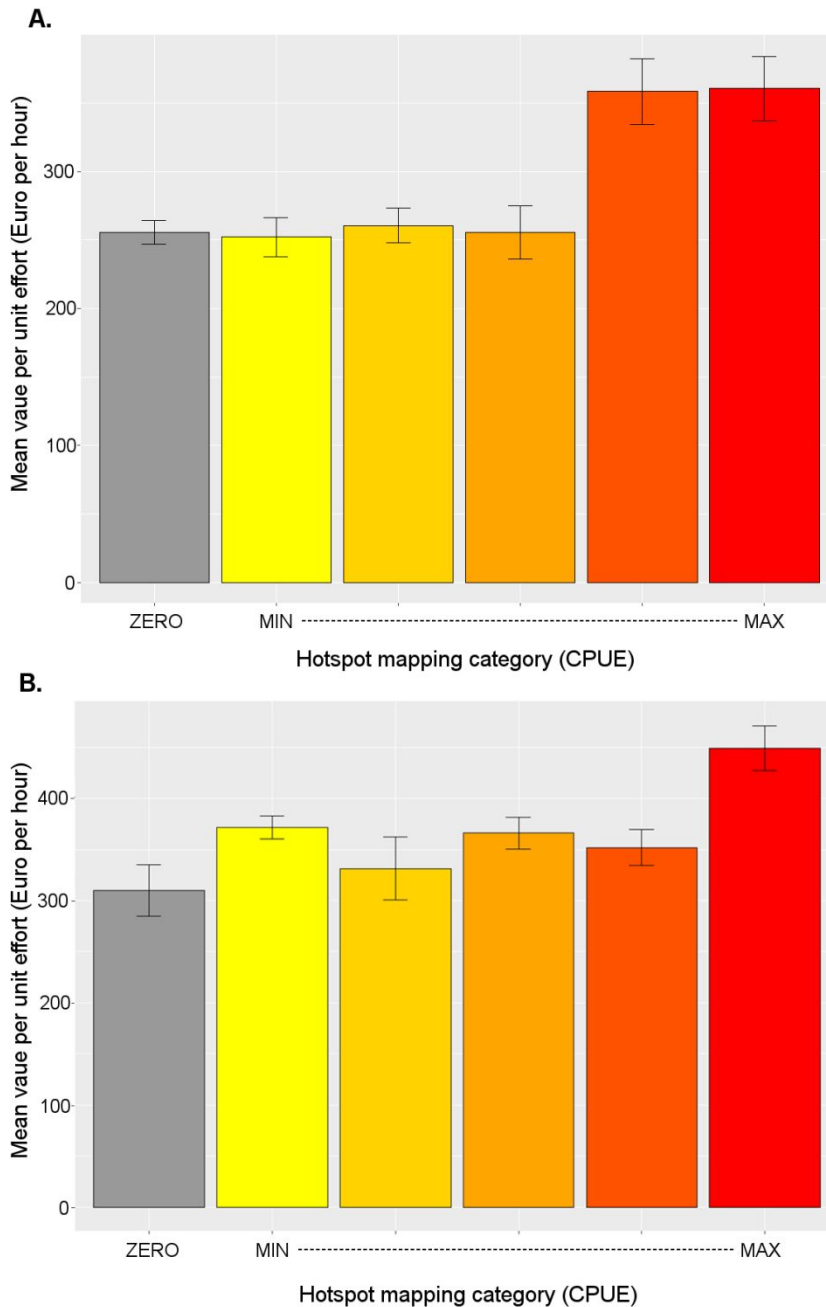
Area	Species	χ^2	df	p
Celtic Sea	Cod	18.12	4	0.001
	Haddock	14	4	0.007
	Hake	26.5	4	<0.001
	Whiting	14.25	4	0.006
	All	33.63	4	<0.001
North Sea & Skagerrak	Cod	19.769	4	<0.001
	Haddock	12.667	4	0.013
	Hake	3.500	4	0.478
	Whiting	8.000	4	0.092
	All	22.154	4	<0.001

Total Catch Hotspots and VPUE

There was no significant difference between the different species and size classes of each when examining the relationship between the mean value of catches in each grid cell and the associated level of consistent catches (based on CPUE) for Irish and Danish fisheries ($\chi^2=4.855$, $df=7$, $p=0.678$ and $\chi^2=5.840$, $df=7$, $p=0.559$);). There was, however, a significant overall relationship between the mean value of catches in each grid cell of the hotspot maps examined and the associated level of consistent catches identified for both Irish and Danish fisheries ($\chi^2=42.395$, $df=5$, $p<0.001$ and $\chi^2=31.273$, $df=4$, $p<0.001$; Figure 54). Post-hoc tests revealed that the value of total catches associated with the highest two CPUE categories in the hotspot maps were greater than the value of catches associated with all other hotspot mapping CPUE categories for Irish fisheries. For the Danish fisheries it was only the

312 highest CPUE category where the value of total catches were greater than the value of catches
 313 associated with all other hotspot mapping CPUE categories.

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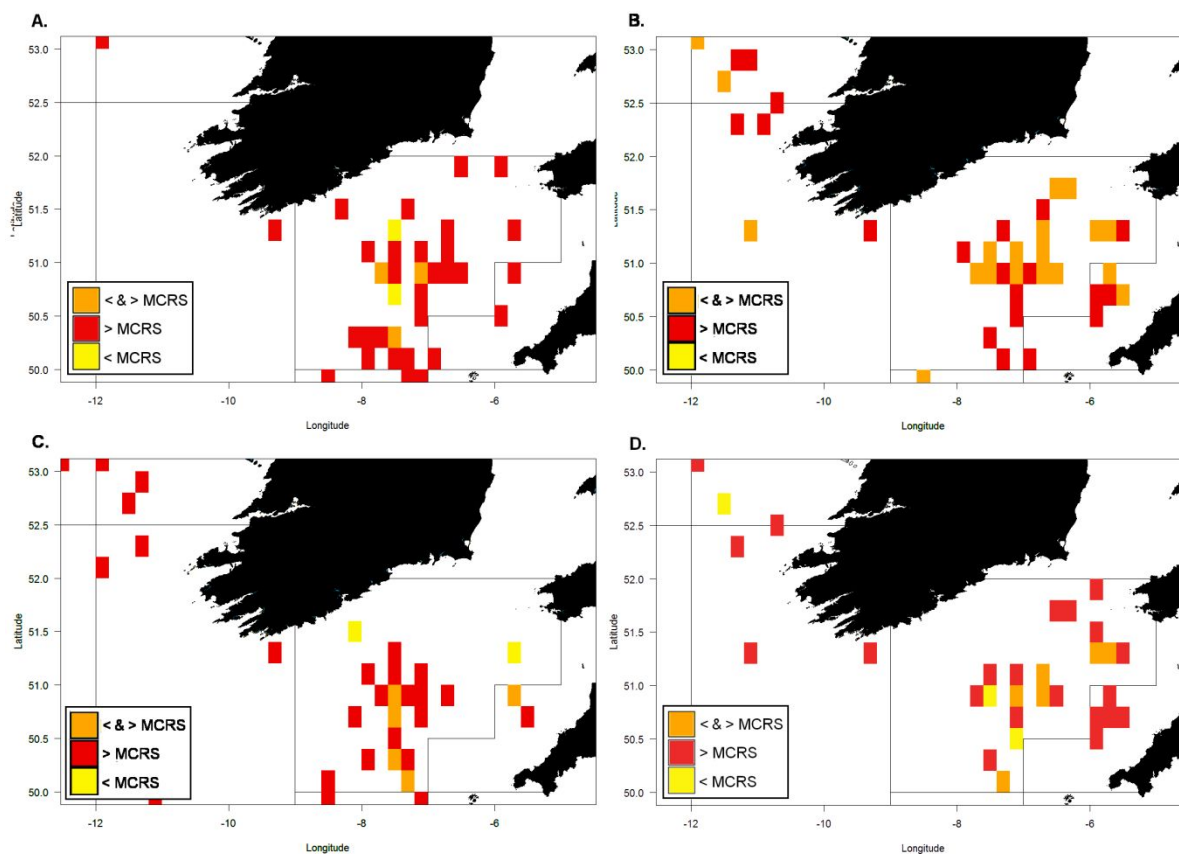


315

316 Figure 54: Bar charts showing the relationship between the mean VPUE of fishing effort in each 0.2 x 0.2 degree
 317 rectangle and the associated levels of catches (CPUE) consistently identified in these areas by hotspot maps
 318 amalgamated for all species and size classes for A. the Irish fishery in the Celtic Sea and B. the Danish fishery in
 319 the North Sea and Skagerrak

320 **Below MCRS Avoidance Scenarios**

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3 321 For all four species in the Celtic Sea (Figure 65) the majority of the grid cells where there are
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5 322 consistently high catch rates for both the < and > MCRS fish are concentrated between 5°W-9°W and
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7 323 50°N-53°N. Apart from haddock, the consistent >MCRS category occupies the most cells for each
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9 324 species for Celtic Sea fisheries. Haddock had an equal number of grid cells assigned to the >MCRS
10
11 325 category and to the overlap of the <&>MCRS category. For all species the percentage of vessel activity
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13 326 in each of these areas in relation to activity in the Celtic Sea study area is greater for areas with
14
15 327 consistently high catch rates of >MCRS species compared to where high >MCRS catch rates overlap
16
17 328 with <MCRS catch rates (Supplementary 23). The highest percentage of vessel activity was recorded
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19 329 in areas with just >MCRS cod, with 10.8% of all fishing activity in the Celtic Sea area in 2016 occurring
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21 330 in this area. For each species there was no statistically significant difference between the mean value
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23 331 of catches attributed to each of the cell categorisation classes (Table 3), although there was a visible
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25 332 trend for lower values being associated with areas of high <MCRS catches for whiting and to a lesser
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27 333 extent cod (Supplementary 23).
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336 Figure 65: Maps showing where areas identified as consistently having high catch rates, as identified by hotspot
 337 maps constructed with observer data from 2010-2015 (top 40% of catches based on kg caught per hour of fishing
 338 activity) for >MCRS and <MCRS catches overlap and where they are spatially separate for A. Cod B. Haddock, C.
 339 Hake and D. Whiting using VMS data from 2016.

340 Table 3: Model results testing the relationship between the category of grid cells (< & > MCRS catches, >MCRS
 341 catches and <MCRS catches) with the mean value of total catches associated with these grid cells for the Irish
 342 fleet operating in the Celtic Sea and the Danish fleet operating in the North Sea and Skagerrak. Significant value
 343 ($p < 0.05$) are indicated in bold.

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Area	Species	χ^2	df	p
Celtic Sea	Cod	2.341	2	0.310
	Haddock	1.749	2	0.417

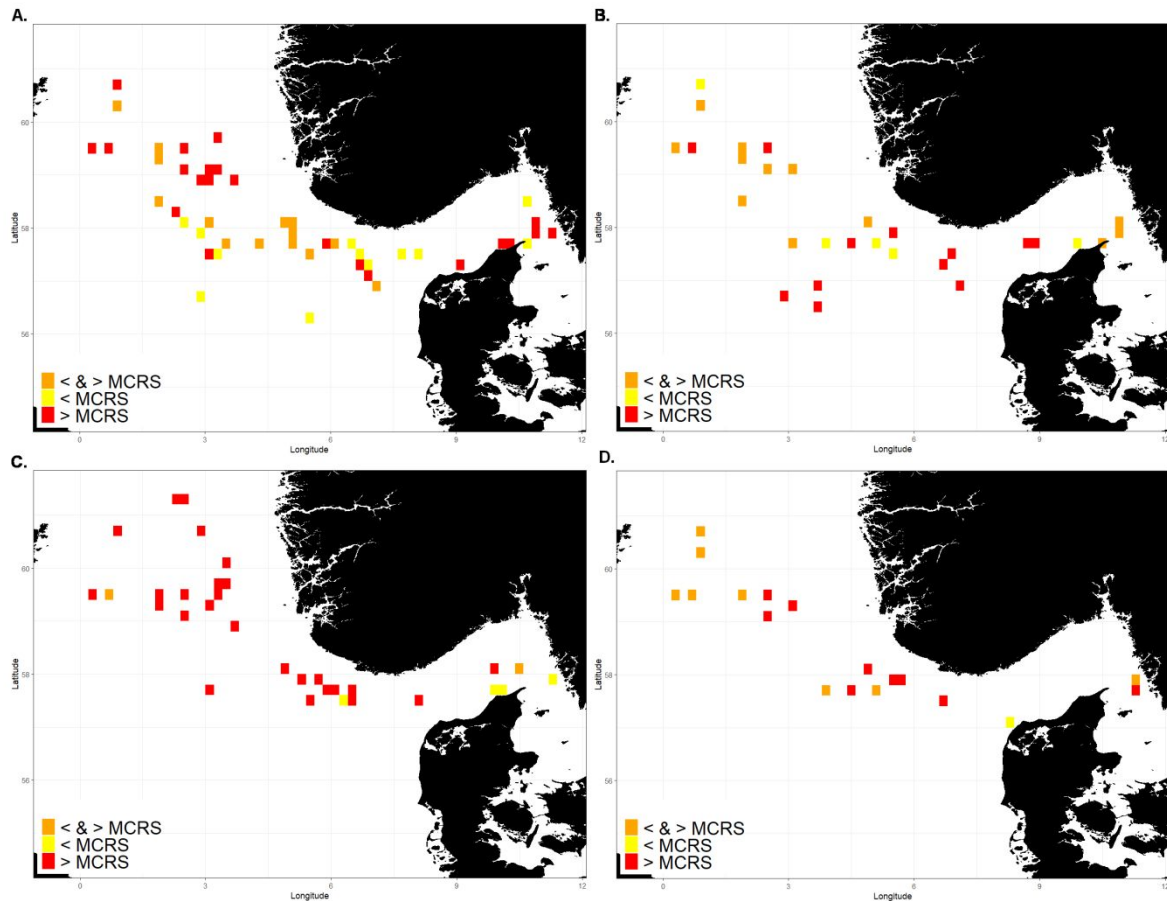
	Hake	0.207	2	0.901
	Whiting	5.355	2	0.069
North Sea & Skagerrak	Cod	4.405	2	0.111
	Haddock	2.596	2	0.2731
	Hake	15.020	2	<0.001
	Whiting	8.090	2	0.018

346

347 For North Sea Danish fisheries, an area between 58°-60° N and 2-4° E is associated with high >MCRS
348 catch rates of cod. There is also a cluster of high >MCRS grid cells in the Skagerrak, but it should be
349 noted that the MCRS is 5 cm lower for cod in Skagerrak than in the North Sea (Figure 76). Above MCRS
350 catch rates of cod are associated with the highest number of grid cells and associated vessel activity
351 but there was no significant difference in value fished between the three categories. Areas identified
352 as having high catch rates of <MCRS cod are spread across the central North Sea and Skagerrak, while
353 the overlap of the two size classes are identified throughout the North Sea. For haddock, 12 cells were
354 assigned to both >MCRS catch rates and the overlap of > and < MCRS catch rates. A greater proportion
355 of vessel activity was associated with the overlapped areas although there was no significant
356 difference in the mean value of catches across all three categories (Supplementary 23, Table 3). For
357 hake, 26 out of 32 cells were categorised as having high >MCRS catch rates and only 2 cells were
358 identified as having overlapping high < and > MCRS catch rates. In terms of vessel activity, the majority
359 of overall vessel activity also occurred in the >MCRS catch rate areas, although areas associated with
360 high <MCRS catch rates of hake also have a large share of activity while representing a much smaller
361 area. Only a minute part of the vessel activity is associated with the two cells identified as overlapping
362 areas but these two cells have the highest mean VPUE (Supplementary 32). Whiting and hake show
363 the highest ~~discarding~~ hotspot cell counts and vessel activity associated with high >MCRS catch rate
364 areas. However, where overlapping areas were few for hake, the smallest cell count and vessel activity

365 occur for <MCRS associated areas for whiting. Just like for hake, the mean VPUE of cells is highest for
 366 overlapping areas for whiting too (Supplementary 32; Table 3).

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369 Figure 76: Maps showing where cells identified as consistently having high catch rates, as identified by hotspot
 370 maps constructed with observer data from 2010-2015 (top 40% of catches based on kg caught per hour of fishing
 371 activity) for >MCRS and <MCRS catches overlap and where they are spatially separate for A. Cod B. Haddock, C.
 372 Hake and D. Whiting using VMS data from 2016.

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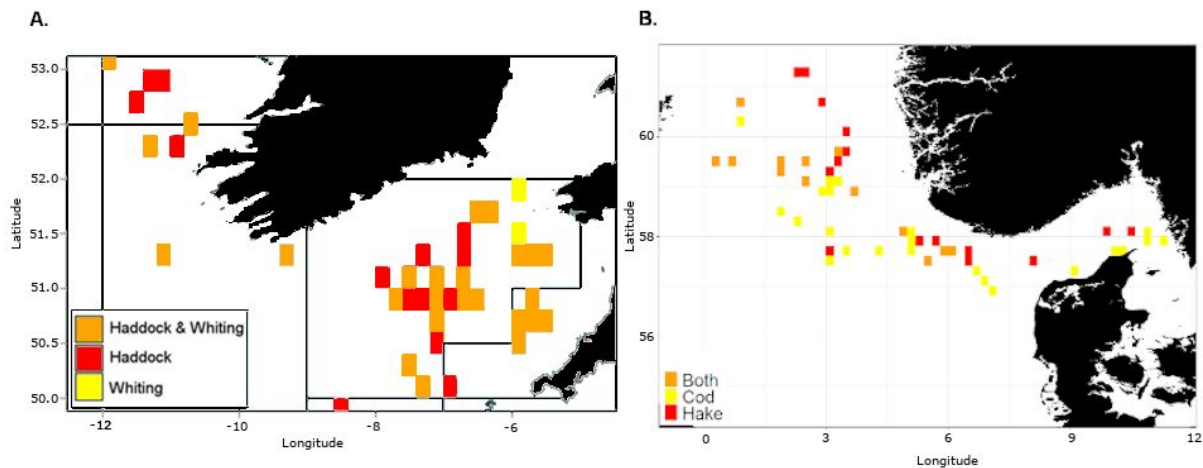
374 Species avoidance scenarios

375 The majority of grid cells in Figure 87A represent areas identified as having consistently high catch
 376 rates of both >MCRS haddock and whiting, with this overlap being identified in 26 grid cells and by

60

377 3.6% of all fishing activity in the region ([Supplementary 4](#)). Only 2 grid cells and 0.6% of the area's
 378 fishing activity are represented by areas with high whiting catches that are not overlapped by those
 379 areas with high haddock catches. There is no statistically significant difference between the mean
 380 value of catches associated with each of the areas identified on the map ($\chi^2=0.198$, $df=2$, $p=0.906$).

381



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383 Figure 87: Map showing where areas identified as consistently having high catch rates, as identified by hotspot
 384 maps constructed with observer data from 2010-2015 (top 40% of catches based on kg caught per hour of fishing
 385 activity) for a. haddock and whiting catches overlap and where they are spatially separate in the Celtic Sea and
 386 B. cod and hake and where they overlap in the North Sea and Skagerrak.

387 In the Danish example cod has the highest number of grid cells assigned as having high catch rates
 388 while cells identified as having high catch rates of hake but no cod amount to 15 cells (Figure 87B). A
 389 total of 13 cells are identified as having overlap between cod and hake, of which 9 occur in the
 390 northern North Sea at around 58°-60.5° N and 0.5-3.5° E. The vessel activity is much more associated
 391 with cod related grid cells, while the vessel activity in grid cells with overlap between cod and hake
 392 and purely hake is roughly the same at 1.8-2.6% compared to 7.9% for areas with cod only
 393 ([Supplementary 4](#)). However, while the 13 cells with high catch shares of both cod and hake have a
 394 low share of the total vessel activity, the mean value per unit of effort is significant different

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3 395 ($\chi^2=10.345$, $df=2$, $p=0.006$) between groups and is highest in the 13 cells with overlap between hake
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5 396 and cod.
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10 11 398 **DISCUSSION**

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14 399 This study sought to determine if the extracted value of fisheries could be maintained following the
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16 400 relocation of fishing effort to avoid unwanted catches. First a link between discarding hotspots and
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18 401 the VPUE of catches was established across two case studies from Irish and Danish fisheries for four
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20 402 key demersal species. In the context of the introduction of the LO we considered how hotspot mapping
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22 403 tools may influence fishing behaviour, again finding a link between areas with predicted high catch
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24 404 rates and subsequent high VPUE of fishing activities. Further exploration of <MCRS and species
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26 405 avoidance scenarios using hotspot maps did demonstrate in some cases the options to avoid <MCRS
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28 406 fish or choke species remain limited. Although options did exist to avoid unwanted catches while still
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30 407 fishing in areas of high VPUE for all examples considered [\(Figure 9\)](#). Spatio-temporal measures,
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34 408 potentially coupled with advanced mapping technology, may prove useful to optimise the use of
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36 409 available quotas. This work did further provide an opportunity to examine how a tool developed for
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38 410 assisting in avoiding unwanted catches the Celtic Sea can be applied elsewhere, demonstrating the
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40 411 wider applicability of this methodology.
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	Irish Case Study	Danish Case Study
Discard hotspots and VPUE data	Significant positive relationship between occurrences of discarding hotspots and total VPUE for cod, haddock, hake, whiting & all species combined.	Significant positive relationship between occurrences of discarding hotspots and total VPUE for cod, haddock & all species combined.
For both there remain areas with high VPUE not associated with discarding hotspots.		
Total catch hotspots and VPUE data	Significant positive relationship between species hotspots and VPUE from total landings in the subsequent year.	
Below MCRS avoidance scenarios	More vessel activity associated with areas where >MCRS catches don't overlap with <MCRS catches for all 4 species. No difference in VPUE associated with each catch hotspot category.	More vessel activity associated with areas where >MCRS catches don't overlap with <MCRS catches for cod, hake & whiting. Higher VPUE values associated with <&>MCRS hotspot overlap areas for hake & whiting
Species avoidance scenarios	Less fishing activity recorded in areas where haddock and whiting hotspots overlap compared to where they don't. Although no differences in VPUE of total catches associated with each area whether hotspots overlap or not.	More vessel activity associated with high cod hotspots than high hake hotspots, or the overlap of the two. VPUE of total catches is highest in areas where high hake and cod hotspots overlap.

413

414 Figure 9: A summary of the results of each of the four sections of the analysis for the two case study
 415 areas.

416

417 When developing tools to aid in the spatio-temporal avoidance of unwanted catches it is important to
 418 firstly consider how discarding practices relate to the wider fishery and overall VPUE of fishing
 419 activities. Our results show that discarding hotspots are widespread across each study region for all
 420 four species considered, but there is clear spatial variation in discarding between each of the species.

421 We have considered these spatial distribution patterns broadly on a multi-annual basis, and more
 422 nuanced patterns may be evident if data were explored on a more seasonal basis. Due to the limited
 423 nature of observer data used in the construction of the hotspot maps, however, we were only able to
 424 construct an annual overview of fishing patterns in relation to discard and catch hotspots. There is
 425 also the potential that discarding trends may vary on shorter time scales and be influenced by the time

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3 426 left before a boat returns to port or by quota availability at the end of a season, although contradictory
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5 427 evidence exists as to whether remaining quota allocation influences discarding behaviour (Poos *et al.*,
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7 428 2010; Calderwood and Reid, 2019). Again the resolution of the data available did not allow further
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9 429 consideration of these factors at this time.
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12
13 430 Still, Aas may be expected in most instances the discarding hotspots are found in areas where there
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15 431 are frequent occurrences of each species (Gerritsen *et al.*, 2012; Marine Institute, 2018). Where
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17 432 discarding hotspots occur in areas not associated with high catch rates of the same species they are
18
19 433 often associated with areas of high catches of other commercial species. For example discarding
20
21 434 hotspots are associated with ICES functional unit 22, also known as ‘The Smalls’, a prawn ground
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23 435 where landings are dominated by *Nephrops norvegicus* in the Celtic Sea (ICES, 2018a), and hotspots in
24
25 436 the Skagerrak bordering the Kattegat management area which are also important fishing grounds for
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27 437 *Nephrops* (Ravensbeck *et al.*, 2015; ICES, 2018b). There are also a few discarding hotspots for cod and
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29 438 haddock occurring in the Central North Sea, likely associated with plaice catches in these shallower
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31 439 areas (Mortensen *et al.*, 2015; ICES, 2018c). For discarding hotspots that occur in areas where there
32
33 440 are high catches of other commercially important species, such as *Nephrops*, spatial avoidance may
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35 441 not be possible as a tactic to reduce discards. In such areas increased uptake of more selective gears,
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37 442 that may allow for the escape of gadoid species while retaining flatfish and crustaceans could be
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39 443 beneficial instead (O’Neill *et al.*, 2019). Identifying these discarding hotspots that are primarily the
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41 444 result of bycatch in other species fisheries is of critical importance and this could be achieved by
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43 445 focussing similar analyses on more fleet segments and metiers, but this again requires a greater
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45 446 provision of fisheries data from the fleet. ~~It is particularly obvious in the *Nephrops* cases we identified.~~
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47 447 ~~These fisheries have been the subject of considerable selectivity research (e.g. Cosgrove *et al.*, 2018),~~
48
49 448 ~~and are the most promising for a gear based discard reduction. Gear based selectivity measures may~~
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51 449 ~~be less promising in the case of mixed demersal gadoid fisheries, unless combined with other~~
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53 450 ~~measures (Sigurrdottir *et al.*, 2015), and this should include behavioural and tactical measures (Pointin~~
54
55 451 ~~*et al.*, 2019).~~

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3 452 If a fisher is a profit maximiser, and economic drivers influence the spatial nature of fisheries, the
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5 453 results from this study indicate that there are options to fish in areas associated with higher value
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8 454 catches while potentially avoiding catches that would otherwise be discarded. ~~There is~~
9
10 455 ~~understandably~~ greater value associated with cells that have consistently high catches of all species
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12 456 studied. ~~This is not unexpected, with~~ Certainly larger volumes of landings are often ~~being~~ associated
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14 457 with higher earnings in mixed fisheries, ~~regardless of species caught~~ as long as there are a good mix of
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16 458 marketable species within the catch. We do, however, only consider the value of landed catches and
17
18 459 not the total profits made by a vessel after accounting for running and staff costs during this study.
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20 460 Travelling further to areas where there is the potential to catch more valuable species or larger
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22 461 volumes of fish may be counteracted by increased fuel costs associated with steaming to these
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24 462 locations. Yet the majority of discarding hotspots are not associated with areas with the highest value
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26 463 catches suggesting fishers ~~can~~ have many potential options to make money across both case study
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28 464 areas while avoiding unwanted catches.

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33 465 Fine scale spatial knowledge of fisheries and the economic value associated with them may be
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35 466 important in helping to provide the knowledge to reduce unwanted catches while maintaining income
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37 467 (Mateo *et al.*, 2017; Calderwood *et al.*, 2019). Developing tools and resources to effectively
38
39 468 communicate such information to the fishing community could therefore be essential in both of the
40
41 469 case study areas to ensure fishers are enabled to make optimum decisions with regard to where to
42
43 470 fish to optimise catches while maintaining profitability. This could take the form of fleet
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45 471 communication programs, for which numerous examples ~~are increasingly being~~ have been highlighted
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47 472 from fisheries throughout the world (Gilman *et al.*, 2006; Eliassen and Bichel, 2015; Little *et al.*, 2015),
48
49 473 or decision support tools provided by the scientific community as are increasingly being developed.
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51 474 ~~Decision support tools provided by the scientific community, either using survey or commercial data,~~
52
53 475 ~~could also be beneficial. Again such tools are increasingly being developed but our results indicate the~~
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55 476 ~~real potential of such applications in information decision making in Irish and Danish fisheries~~ (Reid *et*
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57 477 *al.*, 2019).

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3 478 It is worth noting however, that in the Irish case study many of the high VPUE areas coincide with high
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5 479 value catches of anglerfish and megrim along the 12°W line to the south west of the Celtic Sea in
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7 480 addition to the *Nephrops* fishing grounds (Marine Institute, 2018). Thus although it is possible to fish
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9 481 in areas where high revenues can be made while avoiding discarding hotspots, these high revenue
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11 482 areas may not correspond to areas of high catches of all species. Certainly if quota for these other high
12
13 483 value species were exhausted Irish fishers may not have the same opportunities to target areas with
14
15 484 known high value catches while reducing unwanted catches. For this case study area there are
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17 485 certainly trade-offs that would have to be made with regard to maximising revenues while avoiding
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19 486 low quota or quota restricted species.
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24 487 Despite general similarities in the relationship observed between the discarding patterns and overall
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26 488 VPUE in the two case studies in 2010-2015, and without considering how VPUE values may be driven
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28 489 by different species in each area, the strategies adopted to avoid unwanted catches may be quite
29
30 490 different between Danish and Irish fishers. This is mainly due to the difference in the quota allocation
31
32 491 and management system in these two countries. Danish fishers may better be able to mitigate quota
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34 492 restriction by acquiring additional quota for the specific species since Denmark manages its quota
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36 493 allocations with an ITQ system (Andersen *et al.*, 2010; Mortensen *et al.*, 2018), while Ireland has fixed
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38 494 monthly quotas (DAFM, 2016; Calderwood and Reid, 2019). With the LO in place, fisheries may face
39
40 495 four categories of choke situations, which can close the fishery. These are categorised by the North
41
42 496 Sea Advisory Council NSAC as (North Sea Advisory Council, 2017):
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47 497 **Category 1: Sufficient quota at Member State level—choke is due to distribution within the Member**
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49 498 **State such that a region or fleet segment does not have enough and this can be resolved by the**
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51 499 **Member State itself.**
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55 500 **Category 2: Sufficient quota at EU level, but insufficient quota at MS level—choke is due to a mismatch**
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57 501 **of catches and the distribution of quotas between Member States and can theoretically be resolved**
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59 502 **between themselves in a regional context.**
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3 503 ~~Category 3: Insufficient quota at EU level—choke is due to insufficient quota within the relevant sea~~
4 ~~basin to cover present catches or catch levels that can be realistically reduced, resulting in a total stop~~
5 504 ~~of fishing for a Member State or Member States.~~
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10 506 ~~Category 4: Economic choking may occur at the vessel level when there is a considerable bycatch of a~~
11 ~~low value species and the boat is filled with fish that will not deliver a profit~~
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16 508 Unlike Irish fishers, Danish fishers can therefore avoid ~~the~~ Category 1 choke situations, as defined by
17 the North Sea Advisory Council, whereby there is sufficient quota at a member state level but a choke
18 509 is caused by the way in which quota is distributed within a region or fleet segment (North Sea Advisory
19 Council, 2017). However, depending on the price of acquiring additional quota, this may induce a loss,
20 510 as quota prices change according to supply and demand (Mortensen *et al.*, 2018). While the ITQ
21 511 system allows more flexibility for Danish fishers, this ~~doesn't~~ does not mean that Danish fishers may
22 512 simply trade their way out of quota restrictions and consideration of such costs in relation to the value
23 513 of landed catches is important, ~~both because of profitability of the fishery and because of the~~
24 514 ~~remaining choke categories. Additionally there may be insufficient quota at member state or EU levels~~
25 515 (choke categories 2 & 3 (North Sea Advisory Council, 2017)).
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39 518 Regardless of the flexibilities available within different quota management systems around the EU the
40 519 introduction of the LO will require fishers to increase efforts to avoid unwanted species and
41 520 subsequent choke situations (Borges and Penas Lado, 2019). This legislation presents a pressing need
42 521 for fishers to avoid catches <MCRS, ~~as prior to the introduction of the LO these had to be discarded~~
43 522 ~~but now they have to be landed and count against available quota, despite some limited *de minimis*~~
44 523 exemptions, as these now count against quotas without providing the revenue associated with >MCRS
45 524 landings (European Commission, 2013). The mapping methodology presented here can highlight areas
46 525 with consistently high catches of both the < and >MCRS components of catches, indicating where
47 526 there is a higher chance of catching larger individuals while avoiding juveniles, and hence maximising
48 527 revenue. But options to avoid certain components of the catch are difficult for some species. In the
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3 528 Celtic Sea for example there are no areas where cells representing <MCRS haddock hotspots are not
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5 529 overlapped by >MCRS hotspots. Cells associated solely with >MCRS hotspots have, however, been
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8 530 identified with close proximity to all of the haddock hotspots and with similar VPUE ~~associated with~~
9
10 531 ~~all of these cells values~~. There is, therefore, no economic reason to not avoid the overlap areas. ~~But if~~
11
12 532 all vessels chose to adopt this behaviour, ~~however~~, the displacement of fishing activity could alter the
13
14 533 associated fleet economics in the future. This is also true of the Danish fishery, where although the
15
16 534 value from fishing in all of the haddock hotspots is fairly consistent a larger proportion of fishing
17
18 535 activity is currently associated with the overlap hotspot areas. Such overlap, as seen in both case
19
20 536 studies, is likely due to the frequent co-occurrence of adult and juvenile fish with the whole catch
21
22 537 composition having an influence on fishing behaviour. This could again lead to displacement of fishing
23
24 538 activities and increased pressure on the >MCRS hotspots if vessels were to use the hotspot maps in
25
26 539 an effort to avoid juvenile haddock catches.

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31 540 ~~Again~~ These examples highlight how the provision of such mapping tools to the fishing industry could
32
33 541 be useful in reducing <MCRS catches. More modelling work, potentially utilising Random Utility
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35 542 Models, could be of use for both the Irish and Danish case studies presented here to determine how
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37 543 relocation of fishing effort may further affect fleet economics (Bastardie *et al.*, 2014). To do this a full
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39 544 understanding of the drivers of fishing behaviour is required, which would again require the provision
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41 545 of more catch and effort data. In the Danish fishery for example high catches of <MCRS hake occur in
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43 546 areas with low overall associated VPUE. This would provide an incentive to avoid these areas and
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45 547 target the >MCRS only areas where greater revenues can be achieved. However, the main driver for
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47 548 discarding of hake in the Danish demersal mixed fishery is likely not undersized catches, as by far the
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49 549 majority of observed discards of hake has previously been shown to be >MCRS, ~~with~~ low market
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51 550 value and damaged fish are therefore being pointed to as more important discard drivers (Catchpole
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53 551 *et al.*, 2018; Plet-Hansen *et al.*, 2019). As such it may be more important for Danish fishers to avoid all
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55 552 areas associated with high hake catches and concentrate on catching other species with a higher
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3 553 deemed value. This kind of understanding of the dynamics at play in the decision making process is
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5 554 important for managers to consider when providing appropriate information to fishers.
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9 555 ~~Like the avoidance of <MCRS fish, choke situations are a key concern for fishers operating under the~~
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11 556 ~~LO, especially for mixed fisheries where one species can be more quota restricted than another co-~~
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13 557 ~~occurring species (Schrope, 2010).~~ Greater differences were apparent between the species and choke

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15 558 avoidance scenarios compared to the <MCRS avoidance scenarios for each case study. There is less
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17 559 overlap of cod and hake in the Danish fishery than for haddock and whiting in the Irish fishery,
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19 560 potentially providing more opportunities for the Danish fishery to avoid a choke. However, there is a
20
21 561 higher VPUE associated with overlapping areas for cod and hake in the North Sea and Skagerrak and
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23 562 because of the potential discard drivers for hake, it may be that the Danish fishery has an economic
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25 563 driver for risking catching hake while targeting cod. Managers should maybe consider how they could
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27 564 further incentivise fishing in areas where hake catches can be minimised. In the Irish example there is
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29 565 no difference in the value of catches in the three areas highlighted (single species or overlap) so no
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31 566 financial deterrent from avoiding hotspots where both whiting and haddock are commonly caught
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33 567 together. Further, with very few areas where whiting hotspots occur alone there could potentially be
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35 568 a displacement of vessels into the whiting only areas if vessels were to try and reduce the risk of also
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37 569 catching haddock. Again this might alter whiting catch rates and the overall VPUE associated with
38
39 570 these areas in the long run. This emphasises that while there are some economically attractive areas
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41 571 to fish to avoid unwanted catches of a given species in the Celtic Sea the generally homogenous nature
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43 572 of the fishery makes it is difficult to do this without unintended impacts elsewhere or on other species.

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45 573 The overall similarities in findings between the case studies despite the difference in both observer
46
47 574 coverage and management framework gives us hope that the methods applied do reveal useful and
48
49 575 pertinent spatial patterns. Better understanding of the relationship between discarding hotspots and
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51 576 fleet economics can help us to further understand the drivers of fishing behaviour in space and time.

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53 577 ~~And w~~hen it comes to avoiding <MCRS fish in particular there are numerous options across both our
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3 578 case study examples to do so ~~through the use of spatial avoidance~~, while still fishing in areas associated
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5 579 with high VPUE. The tools highlighted in this study also offer potential in choke situations, but the
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7 580 nature of mixed fisheries mean the options to fish perfectly in line with available quotas are limited.
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9 581 A further barrier to operational use of this type of analysis lies in the very limited amount of observer
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11 582 data collected, at less than 1% of fishing operations, and the risk of an observer effect influencing this
12
13 583 data (Schaeffer and Hoffman, 2004; Plet-Hansen *et al.*, 2018). Certainly the variation in observer
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15 584 coverage on different types of vessels, in different areas and across different seasons and years may
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17 585 influence the resultant hotspot maps. With the data available to us, however, pooling data annually
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19 586 provided the most useful information with regard to better informing fishing behaviour to avoid
20
21 587 unwanted catches. Because of this relative paucity of data, the hotspot analyses had to be done using
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23 588 data pooled over a number of years. Yet if we were able to access accurate and extensive catch data
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25 589 from the fishing fleets, the analyses could be more up to date, more resolved in time and space, and
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27 590 more useful to the fishers themselves. This view was expressed by a number of fishers who had
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29 591 considered using the hotspot maps themselves and would make the tool much more valuable.
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35 592 The possibility of adding economic data and information into such tools could also be beneficial in
36
37 593 further informing decision making in fisheries and fisheries management. It is however recognized
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39 594 that this would require a sea change in the way fisheries currently provide data, but the pay-offs could
40
41 595 be substantial. Certainly the mapping tool presented, among a suite of those currently being
42
43 596 developed, could play an important role in providing the spatial solutions to reduce unwanted catches.
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45 597 This spatial data could also highlight areas in which the use of more selective gears is most needed.
46
47 598 But further consideration needs to be given to how ~~the information presented in such maps such~~
48
49 599 information is related to the potential value of catches, and how this can be effectively communicated
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51 600 so that fishers can make the most informed decision on how to optimise catches while remaining
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53 601 profitable. relate to fleet economics and other drivers of fishing behaviour, and how such economic
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55 602 data could be provided and presented to determine if the advice presented is likely to be followed.
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59 603 Additionally, fisheries scientists need to avoid giving advice on fishing behaviour for the industry that

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3 604 is economically unviable. Further incentives may be required in some instances to either utilise more
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5 605 selective gears or avoid areas with high bycatch risk. The ability to identify where these measures may
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8 606 be required, ~~however~~, is a useful tool in the management of fisheries.
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11 607

12 13 14 608 **Acknowledgements**

15
16
17 609 This work has been funded by the European Union's Horizon 2020 research and innovation
18
19 610 programme under Grant Agreement DiscardLess No 633680.

20 21 22 611 **Data Availability Statement**

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25 612 The data underlying this article cannot be shared publicly due to its' sensitive nature on a commercial
26
27 613 and personal level and the requirements of GDPR legislation. It may be possible to share amalgamated
28
29 614 data on reasonable request to the corresponding author.

30 31 32 33 615 **Author Contributions**

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36 616 J.C. and K.P-H. contributed equally to the development of the work presented, the analysis conducted
37
38 617 and the writing of the manuscript. C.U. and D.R. guided the development of this work and contributed
39
40 618 to the writing of the manuscript.
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