

A comparison of VMS and AIS data: the effect of data coverage and vessel position recording frequency on estimates of fishing footprints

Shepperson, Jennifer; Hintzen, Niels T.; Szostek, Claire; Bell, Ewen; Murray, Lee; Kaiser, Michel

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1	A comparison of VMS and AIS data; the effect of data coverage and vessel
2	position recording frequency on estimates of fishing footprints
3	
4	Jennifer L. Shepperson ^{1,3} , Niels T. Hintzen ² , Claire L. Szostek ¹ , Ewen Bell ³ , Lee G. Murray ¹ , Michel J.
5	Kaiser ¹
6	¹ School of Ocean Sciences, Askew Street, Menai Bridge, Anglesey, LL59 5AB
7	² Wageningen Marine Research, PO Box 68, 1970AB IJMUIDEN
8	³ Centre for Environment, Fisheries, and Aquaculture Science, Pakefield Road, Lowestoft, Suffolk, NR33 0HT
9	Corresponding Author: Jennifer L. Shepperson, j.shepperson@bangor.ac.uk, Tel: +44 1248 388501
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21 directly temporally matching data (i.e. only for segments of time in which there were VMS and

- 22 AIS records for a vessel), the time and extent of fishing was very similar between the two types of
- 23 data, although a slightly greater amount of fishing effort (+2.6% fishing hours) was still

24	determined by VMS data compared to AIS data. Using a cubic Hermite spline interpolation of VMS
25	data provided the greatest similarity in the extent of fishing predicted by directly matching VMS
26	and AIS data, but the scale at which the data were analysed (i.e. size of the grid cells) had the
27	greatest influence on the estimates of fishing extent and intensity compared to the data type
28	(AIS, VMS) and data treatment (point, straight line interpolation, or cubic Hermite spline
29	interpolation).
30	The present gaps in coverage of AIS may make it inappropriate for absolute estimates of fishing
31	extent and intensity. VMS already provides a means of collecting more complete fishing position
32	data, shielded from public view. Hence, there is a clear incentive to increase the VMS poll
33	frequency to calculate more accurate fishing footprints, which would ultimately benefit both
34	fishers and scientists.
35	
36	
37	Key Words: vessel monitoring system; automatic identification system; fisheries; footprint;

38 extent; scallop dredging

1. Introduction 40

41

1.1. We need to understand fishing footprints to understand fishing impacts 42

43 Physical disturbance by mobile bottom contacting fishing gears is the largest cause of human 44 disturbance to continental shelves in all areas of the world (Foden et al., 2011). In order to 45 understand the extent and consequences of these disturbances it is necessary to have an 46 accurate understanding of the distribution in space and time of that disturbance. For these 47 reasons, the use of vessel tracking data to analyse patterns of fishing effort and the impact of 48 fishing pressure on marine environments is a key area of fisheries science (Mccauley et al., 2016; Russo et al., 2016; Hintzen et al., 2012; Lee et al., 2010).

50

49

1.2. VMS data is increasingly used to analyse fishing activity, but has 51

limitations 52

53 Satellite based Vessel Monitoring Systems (VMS) were introduced as an enforcement tool, but 54 the resulting data are increasingly important for scientific research and management (Murray *et* 55 al., 2011, 2013; Lambert et al., 2012). Despite the importance of these data, the temporal 56 resolution of VMS is relatively low in Europe usually with a 1 or 2 hourly poll rate. This poll rate is designed as a compromise between adequate resolution and costs to fishers. Interpolation of 57 58 VMS data is typically used to fill in the gaps between successive VMS points to produce a 59 continuous track. VMS data can be joined to grids to analyse the extent of, and patterns in fishing 60 intensity, either as raw point data, or as interpolated tracks. However, the methodology used to 61 analyse VMS data can influence the estimates of fishing intensity (Piet and Hintzen, 2012), and 62 the relationship between fishing intensity and epifaunal biomass (Lambert et al., 2012). In

particular, the grid cell size used for analysis influences the intensity estimates (Hinz *et al.*, 2013;
Lambert *et al.*, 2012; Piet and Quirijns, 2009; Dinmore *et al.*, 2003).

65

AIS data has a higher temporal resolution, and could be used to
 investigate fishing activity.

68 There has been a recent increase in interest in the potential for using publicly available Automatic 69 Identification System (AIS) vessel tracking data to investigate fishing activity (Mccauley et al., 70 2016; Russo et al., 2016; Natale et al., 2015). AIS data is openly available to the public, at high 71 resolution, whilst VMS data is subject to strict confidentiality regulations, which can mean only 72 highly aggregated data is available to scientists outside regulatory bodies (Hinz et al., 2013). High 73 resolution analyses of VMS readily demonstrate the problems and limitations of using aggregated 74 data (Hinz et al., 2013). In order to obtain raw VMS data, non-governmental scientists need to 75 approach each vessel individually, which is impractical to get coverage of the whole UK fleets, and 76 makes extension to the European wide fleet unfeasible. While AIS data may provide a useful 77 alternative source of information with which to understand patterns in fishing activity, the 78 present study reveals the many problems with interpreting AIS data due to gaps in the data. 79 VMS is mandatory on fishing vessels >12m in length in the European Union for enforcement 80 purposes (EC, 2009), whilst AIS is required on vessels >15m for safety purposes. Nevertheless, 81 aspects of the AIS technology, and legislation, mean fishing activity may not be completely 82 recorded by AIS (McCauley et al., 2016). Thus, whilst access to VMS data is subject to 83 confidentiality issues that degrades its utility for research purposes (Hinz et al., 2013), AIS data 84 has different disadvantages, as it can lack consistent coverage. AIS signals are recorded in a 85 different way to VMS data, in that they are broadcast omni-directionally and can be picked up by 86 receivers on land, or by other vessels, as the system was designed to reduce collisions and offer

87 safety mainly when near other traffic or near ports. If a vessel is out of reach of a land based 88 station, the signal must be transmitted from vessel to vessel until it reaches a land station. In 89 areas with relatively low vessel densities, this could cause gaps in coverage. Signals can also be 90 'lost' in areas of very high density traffic. In addition, skippers can turn down the power on the 91 AIS, which reduces the range of the signal, further increasing the likelihood of gaps in coverage. 92 McCauley et al., (2016) argue that having an AIS system on board a vessel, but failing to use it 93 properly, should no longer be viewed as legal compliance. It is also possible for skippers to falsify 94 AIS data, and provide incorrect vessel IDs (McCauley et al., 2016), with the vessel identity of AIS 95 signals not subjected to the same validation process by inspection agencies as VMS data. Despite the positional accuracy of AIS data being comparable to VMS data, there can be considerable 96 97 variation in spatial coverage between different fleets of vessels (Russo et al., 2016).

98

99 1.4. High resolution AIS data might improve footprint estimates of scallop
 100 dredging

101 Besides the lower overall fleet coverage of AIS data (i.e. number of vessels with AIS), it would be 102 useful to understand more about the differences in inferred fishing activity between VMS and AIS 103 data, where the coverage is concurrent. There can be considerable gaps in AIS data coverage in 104 space and time within fleets (Russo et al., 2016; Natale et al., 2015), but there has been little 105 comparison of the recorded activity by the two data sources specifically on trips where both VMS 106 and AIS were actively transmitting data, or assessment of individual variation in coverage. In 107 situations where AIS is the primary data source available to scientists, e.g. in areas where no, or 108 only highly aggregated VMS data is available, it is important to know how the conclusions drawn 109 from AIS would correlate with those drawn from VMS data. In addition, an analysis of concurrent 110 VMS and AIS data would enable us to understand better the complexity in patterns of fishing that 111 may be missed by VMS data due to the issue of temporal position frequency. The structure of the 112 VMS and AIS data itself is essentially the same, a file with coordinates, speed, heading, and vessel 113 ID, which means the same data processing techniques can simply be applied. Nevertheless, 114 despite using the same processing techniques, there could be differences in the resulting 115 conclusions due to differences in the way the data were generated. Primarily, AIS data is available 116 at a much higher poll frequency than VMS data. Finer scale patterns in activity may therefore be 117 seen with AIS data, for example, using a longer 2 hourly poll frequency in VMS could miss shorter 118 steaming sections between tows, giving the impression of long continuous fishing activity, 119 potentially overestimating fishing activity. Alternatively, due to the difficulty in accurately interpolating the tracks between 2 hourly position records, the VMS could also lead to an under-120 121 estimate of the extent and intensity of fishing. 122 Understanding this error becomes particularly important when attempting to understand the 123 environmental footprint of different fishing activities. In this paper, the focus is on scallop fishing, 124 which is considered one of the least environmentally compatible forms of mobile bottom 125 contacting fishing gears (Kaiser et al. 2006). European Union Directives such as the Marine 126 Strategy Framework Directive (MFSD) and Good Ecological Status (GES) use the fishing footprint 127 (spatial distribution of fishing activity) as an indicator of ecosystem health. Understanding the 128 distribution of fishing activity is fundamental to understanding and quantifying the impact that 129 fishing has on the seabed (Kaiser et al., 2016). VMS data provides a means to understand the 130 spatial footprint (extent and intensity) of scallop dredging and other fisheries and this has 131 become a research field of its own (Eigaard et al., 2017; Hinz et al., 2013;Lambert et al., 2012;). 132 However, there is a conflict between the requirement for high temporal and spatial resolution 133 data needed for scientific research, and the publicly available lower temporal and spatial 134 resolution data to uphold confidentiality of commercially sensitive data (Hinz et al., 2012, 135 Lambert et al., 2012). In many trawl fisheries the gear is towed for in excess of four hours and 136

often in a single direction with few deviations. In contrast, scallop dredgers can make short tows

137 (~20 minutes), make tight turns and often tow parallel to their previous tracks, which can make 138 the prediction of trajectories using interpolation methods difficult when the resolution of data is 139 low (Lambert et al., 2012). With scallop dredging activity, higher resolution AIS data may 140 therefore be better able to capture (1) the true footprint of fishing by better capturing the sharp 141 turns made by vessels; and (2) the true fishing effort level, by better capturing the time spent in 142 each activity state (i.e. fishing cf. steaming). This could provide insights into the appropriate 143 treatment of lower resolution VMS data.

Nevertheless, this benefit of higher poll frequency in AIS data may be counteracted by gaps in coverage. Before AIS data can be used as a data source for management, these differences in coverage and interpretation must be understood and addressed. This paper seeks to address this gap in understanding, by comparing the fishing activity of vessels in the English Channel Scallop fishery, on days for which it was possible to obtain both VMS and AIS position records.

149

150 **1.5.** Aims

151 The main aims of this paper were to: (1) determine the relative coverage of AIS data in relation to 152 VMS data at both the fleet and an individual vessel level; and (2) for matching data (from the 153 same vessels in the same time period), determine whether the fishing extent and intensity 154 predicted by three common methods of VMS data analysis (point density, straight line 155 interpolation (Stelzenmuller et al. 2008), and cubic Hermite spline interpolation (Hintzen et al. 156 2010)) showed a comparable accuracy to the higher poll frequency AIS data. Conclusions were 157 drawn about the accuracy of the two data sources (AIS vs VMS) for estimating fishing extent and 158 intensity, considering the relative coverage of each data type, differences in temporal poll 159 frequency, and the level of spatial aggregation used during data analysis.

160

161 2. Methods

2.1. Data Coverage

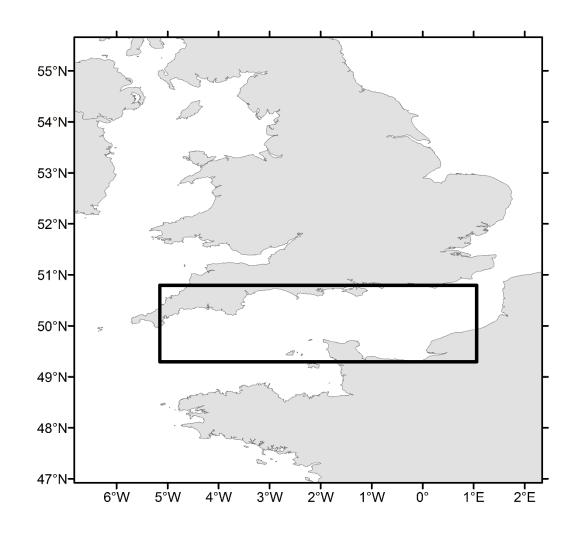
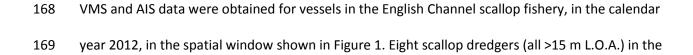


Figure 1. The spatial window in which the VMS and AIS records were recorded. The actual
positions of data points are concealed for confidentiality.



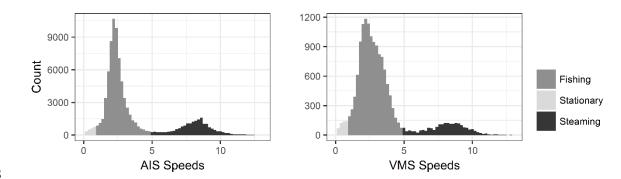
170 English Channel gave permission for their raw VMS data to be used in this analysis. The VMS data 171 included vessel identification data, position, time, speed and heading. AIS position, time, speed, 172 heading, and vessel identification data for the same eight vessels over the same time period were 173 obtained from the company AstraPaging Ltd (http://www.astrapaging.com/), a private AIS data 174 provider. The Maritime Mobile Service Identity (MMSI) field, a nine digit number uniquely 175 identifying a ship radio station installed on each vessel, was used to request the AIS data for the 176 eight vessels. VMS data was provided at a poll frequency of approximately 2 hours, and the AIS 177 data was provided at a poll frequency of approximately 5 minutes.

178 AIS coverage can vary between fleets (Natale et al., 2015), and has also been shown to capture a 179 smaller amount of fishing activity than VMS data (Russo et al., 2016). Therefore, following the 180 initial assessment of data coverage, data were excluded from the analysis if they could not be 181 matched to corresponding VMS or AIS data from that same vessel over the same time period. 182 Initially, for each date, a vessel's VMS data were removed from the analysis if that same vessel 183 had not also recorded AIS data on that date, and vice versa; therefore the term 'comparable date' 184 is used to signify a date on which a particular vessel had recorded both VMS and AIS, which 185 generated 'comparable data'. Nevertheless, even if there were some VMS and AIS for a vessel on 186 a particular day/trip, either dataset may not be complete within the trip. Thus a further category 187 of matching data was identified, by extracting trips where the ratio of the duration of VMS:AIS 188 points was between 0.8 and 1.2, i.e. there was less than 20% mismatch in the duration of VMS compared to AIS, so substantial sections of either data were not missing within a trip. There were 189 190 therefore 2 categories of data: comparable data, which refers to trips for which there is some 191 VMS and AIS for that vessel, but within trip completeness has not been quantified; and matching 192 data, which refers to trips for which the ratio of VMS:AIS is between 0.8 - 1.2, meaning that only 193 a subsection of a trip may be included, which has more complete data from both sources. Only 194 comparable or matching data were used in the comparisons of fishing activity, extent, intensity, 195 and track interpolation.

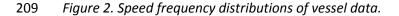
197 **2.2. Data Processing**

198 Both VMS and AIS datasets were subjected to the same data cleaning and processing strategy, 199 using the VMStools packages in R (Hintzen et al., 2012). Duplicate VMS records and records close 200 to (within 1km of) port were removed, along with erroneous position records allocated to land 201 (Lee et al., 2010). Following examination of the frequency distribution of the recorded speeds, 202 position records between 1 – 5knots were classed as fishing activity (Figure 2); these fishing 203 points were either analysed as raw point data, or interpolated to reconstruct higher resolution 204 estimates of the fishing tracks (Section 2.3). The level of data retention of VMS and AIS data was 205 recorded at each stage of data cleaning and processing, to identify differences and similarities in 206 the data, and identify any substantial loss of data.

207







210

211 2.3. Interpolation of tracks

- 212 VMS and AIS data can be analysed in the raw point data format, or vessel tracks can be
- 213 reconstructed using a straight line (SL) interpolation, or a cubic Hermite spline (cHs) interpolation

214 (Hintzen et al., 2012). Succinctly, the cHs method uses information on vessel position, heading 215 and speed at times t and t+1 to define a trajectory. The combination of speed and heading are 216 represented by vectors, and vector length is multiplied by a parameter *fm* that influences the 217 curvature of the interpolations (Lambert et al., 2012). The VMStools package in R provides a 218 function for fm parameter optimisation; the high resolution AIS data was used to determine the 219 optimal fm parameter for cHs interpolation of the VMS data. CHs interpolation of the AIS data 220 was not possible, as there was no higher resolution data for the optimisation process, 221 nevertheless the AIS points were recorded at a high 5 minute poll frequency, so a SL interpolation 222 would give a sufficient level of spatial detail in the tracks. The SL interpolation of the AIS data can 223 be assumed as the most robust estimate of the path taken by the vessels due to its high poll 224 frequency.

225

226 **2.4.** Data Analysis

The number and proportion of points classed as fishing activity were compared between data types and vessels, to identify differences between the data types, and whether these differences varied between individual vessels. To investigate the footprint of fishing activity, points that were classed as fishing activity were joined to spatial grids of 1km, 3km, 5km, and 10km in cell size. These grids were used to calculate the fishing extent and intensity. The interpolated tracks were turned into a series of points approximately every 30 seconds along the track, to analyse in the same way as the raw point data.

The extent of fishing was calculated by counting grid cells which had at least 1 fishing point in them using each data type and interpolation method. The intensity of fishing was compared by calculating the area swept in each grid cell, by summing the area swept per point in the cell, using

237 each data type and interpolation method. Area swept per point was calculated as:

Area swept (km²) = speed (km/h) * time fishing (h) * total dredge width (km) where the total dredge width is assumed to be 0.018km for all vessels, corresponding to the width of 24 individual dredges each measuring 0.75m across. In this analysis a fixed dredge width was used as the actual dredge width was unknown. Total dredge width has been shown to increase with vessel length (Eigaard et al., 2016), but as the estimates of fishing effort in this analysis are relative and not absolute (as only a subsection of vessels were included), a fixed dredge width was considered appropriate.

In addition, the swept area ratio (SAR) was calculated for each grid cell (Gerritsen *et al.,* 2013).

246 The SAR indicates what proportion of the cell has been dredged at least one time, calculated as:

247 Swept area ratio = Area swept (km²) / area of cell (km²)

248 In this study, one year of AIS and VMS data were used, therefore these SAR values relate to a one

249 year time period. A SAR of 1 therefore indicates that on average each part of a grid cell has been

250 dredged one time over the year, a SAR of 2 indicates the whole cell has been swept twice, a SAR

251 of 0.5 indicates that on average half of the cell has been dredged one time. Each VMS

252 interpolation method (point data, SL interpolation, and cHs interpolation) was compared to the SL

AIS interpolations (assumed as the truest fishing tracks) and AIS point data.

254

255 **2.5.** Data Confidentiality

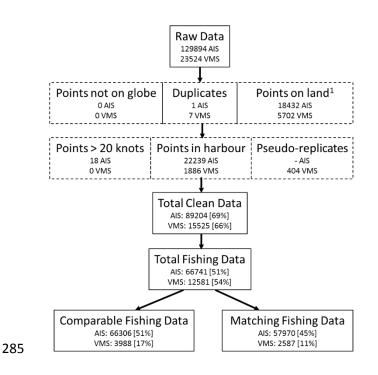
Vessels that contributed to this study are anonymous throughout the analysis, as VMS and AIS data are commercially sensitive, and therefore confidentiality is an important issue. VMS data were provided by fishermen under the condition that the location of fishing activity would not be displayed, therefore the spatial reference has been removed from any maps. The same level of confidentiality has been afforded to the AIS data.

262 **3. Results**

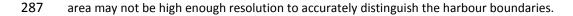
3.1. How do the basic VMS and AIS datasets compare?

The eight studied vessels recorded 129894 AIS points, and 23524 VMS points during the calendar year 2012. After cleaning, there were 89204 AIS and 15525 VMS points remaining (Figure 3) representing a 69% and 66% retention respectively. 51% of the raw uncleaned AIS data and 54% of the raw uncleaned VMS data were retained as representing fishing activity, comprising 66741 and 12581 points respectively. This represented 75% of the clean AIS data and 81% of the clean VMS data.

270 For the comparable trip data (i.e. records which had corresponding AIS or VMS for that vessel on 271 that day) there were 66306 AIS points and 3988 VMS points from seven vessels (thus one vessel 272 was excluded from further analysis). This retained 51% of the raw AIS data and 17% of the raw 273 VMS data, or 99% of the cleaned AIS data and 32% of the cleaned VMS data that represented 274 fishing points. For the matching data (i.e. only trips with a ratio of VMS:AIS within the threshold 275 0.8 – 1.2, to reduce missing data within trips), this was reduced to 57970 AIS points and 2587 276 VMS points (45% and 11% of the raw data, 65% and 17% of the clean data, respectively). A 277 substantial amount of AIS data were therefore missing, i.e. there were a lot of days for which 278 there were VMS data but no corresponding AIS data, but there were comparable VMS records for 279 almost all AIS records. When the data were reduced further to only trips with a high VMS:AIS 280 ratio, 13% of the comparable AIS data and 35% of the comparable VMS data were removed. This 281 indicates that whilst there were more missing AIS data within trips, there were also missing VMS 282 data within trips. The average time interval between all VMS points was 130 minutes, and 283 between AIS was 13 minutes, but when only fishing points were used, the average time interval 284 between VMS points was 114 minutes, and between AIS points was 5 minutes.



¹Points classed as being on land are often just in harbour, but the GIS map used to define the land/harbour



- 289 Figure 3. Preparation and cleaning of AIS and VMS data. Dashed boxes indicate data that is
- 290 *removed, solid boxes indicate retained data.*

- 3.2. How does the assignment of fishing activity compare between VMS and
- AIS data?
- 294 The raw VMS data indicated 29701 hours of fishing, but the raw AIS data only estimated 7647
- 295 hours of fishing, which constitutes a substantial gap in the coverage of AIS data. Despite using
- data from vessels which have both VMS and AIS on-board, the AIS data only captured 26% of the
- time spent fishing compared to VMS data.
- 298 The proportion of time each vessel spent fishing, steaming, and effectively stationary was then
- 299 compared between the comparable AIS and VMS data for each vessel on each day. Overall, using

comparable trip data, the AIS data indicated a total of 5661 hours fishing by all vessels across the
study period, and the VMS data a total of 7751 hours, suggesting 2090 extra fishing hours with
the VMS than the AIS data. The AIS data indicated a total area swept (calculated as area swept
per fishing point: area swept = speed * dredge width * time) of 469km², and the VMS data
indicated a total area swept of 651km².

For the matching data, the AIS data indicated a total of 4924 hours fishing across the study period, the VMS a total of 5053 hours, suggesting 129 extra hours fishing by the VMS data. The AIS data indicated an area swept of 405km², and the VMS an area swept of 406km². If using all available data, the overall extent of fishing is under-estimated by 74% by AIS data compared to VMS data, using comparable data (i.e. trips for which there is some VMS and some AIS data) the overall extent of fishing is under-estimated by 39% when using AIS data, but if using only data that directly matches in time, the extent of area affected by fishing was very similar.

312

313 **3.3.** How does the data coverage and activity assignment differ between

314 individuals?

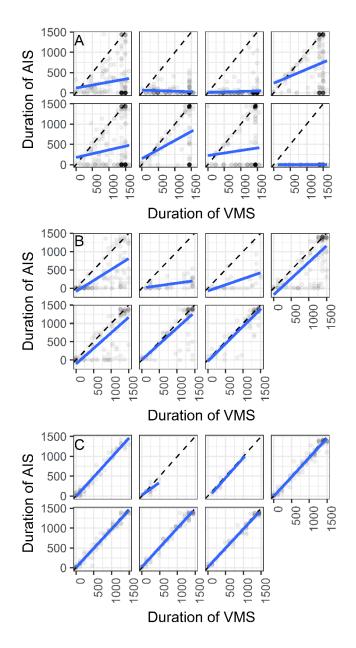
315 A substantial amount of AIS data were missing, however the amount of missing data differed 316 between individual vessels (Table 1). In all cases, more AIS data were missing than VMS data. 317 Thus when retaining only comparable trip data, for some vessels the removal of VMS data was 318 large, e.g. a reduction from 2176 points to 151 points (93% loss); the smallest loss was 34%. In 319 contrast, the greatest loss of AIS data due to having no corresponding VMS data was 2%, and the 320 smallest loss was nil. However, when using only trips with matching data (a high ratio of 321 VMS:AIS), considerably more data were removed; 45-99% of VMS data, and 7-51% of AIS data. 322 This suggests that overall there were substantially more AIS data missing, but there were some 323 trips with considerable amounts of VMS data missing as well.

- 324 The comparable trip VMS data gave a higher estimate of time spent fishing and area swept than
- the AIS data for all individuals (Table 1). The magnitude of this difference varied between 18-92%
- between individual vessels. When using the more closely matched data, the time spent fishing for
- each individual varied from +2% to -26%, and the area swept from +3% to -17%, but the values of
- 328 the absolute difference were relatively low.
- 329
- 330
- Table 1. Number of VMS and AIS points per vessel when using all available data, only comparable
- 332 data, and only matching data. The time spent fishing and area swept columns indicate the total
- 333 for each vessel using comparable and matching VMS and AIS data. The % dif column indicates
- how much smaller the AIS value was than the VMS value.

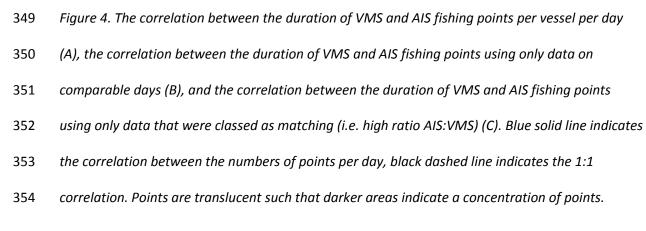
Со	mparak	ole Data														
ID	Number of fishing points (all available)		nts Comparable lost		AIS lost	Time fishing (hours)		% dif	Area swept (km²)		% dif					
	VMS	AIS	VMS	AIS			VMS	AIS		VMS	AIS					
1	1539	7249	616	7227	60%	0.3%	1171	646	-45%	115	64	-44%				
2	1963	942	251	942	87%	0.0%	514	91	-82%	57	10	-82%				
3	2176	1038	151	1036	93%	0.2%	267	96	-64%	25	9	-64%				
4	1581	18086	1023	18075	35%	0.1%	1957	1521	-22%	149	117	-21%				
5	1798	12291	701	12241	61%	0.4%	1328	1043	-21%	100	82	-18%				
6	946	13369	620	13322	34%	0.4%	1263	1120	-11%	105	94	-10%				
7	1921	13766	626	13463	67%	2.0%	1251	1145	-8%	98	93	-5%				
Ma	tching	Data														
ID	Numb	er of	Numb	er of	VMS	AIS	Time f	ishing	% dif	Area s	wept	% dif				
	fishing points		fishing points		fishing points		shing points Matching		lost los	lost	(hours)			(km²)		
	(all available)		Fishing points													
	VMS	AIS	VMS	AIS			VMS	AIS		VMS	AIS					

	VIVIS	AIS	VIVIS	AIS			VIVIS	AIS		VIVIS	AIS	
1	1539	7249	259	5448	83%	25%	497	477	-4%	49	47	-4%
2	1963	942	29	460	99%	51%	58	43	-26%	6	5	-17%
3	2176	1038	40	753	98%	27%	77	69	-10%	7	7	
4	1581	18086	735	16202	54%	10%	1410	1362	-3%	107	105	-2%
5	1798	12291	463	10209	74%	17%	889	869	-2%	67	69	+3%
6	946	13369	521	12464	45%	7%	1042	1048	+1%	87	88	+1%
7	1921	13766	540	12434	72%	10%	1079	1055	+2%	84	85	+1%

337	The correlation between the number of VMS and number of AIS points per day varied between
338	vessels, and depended on the data treatment (Figure 4), but there was a considerable amount of
339	missing fishing data for all vessels. Using comparable trip data only, the correlation between the
340	duration of VMS and AIS fishing records per vessel per day improved significantly. Nevertheless,
341	the duration of AIS data is slightly lower than the VMS data for all vessels. For some vessels there
342	is considerably less AIS data than expected on comparable days, suggesting some gaps in
343	coverage within a trip. The dashed line indicates the 1:1 ratio between VMS and AIS data;
344	matching data were identified as trips that had a ratio of 0.8 – 1.2, therefore matching data
345	showed a strong correlation between the duration of VMS and AIS points by definition (Figure
346	5c).







356 3.4. How does the spatial footprint of fishing compare between VMS and AIS
 357 data?

358

359 3.4.1. Interpolation of VMS and AIS fishing tracks

360 A straight line interpolation of the matching AIS data was used to create the best estimate of the 361 vessels' tracks. The matching VMS data was interpolated using both the straight line (SL) and 362 cubic Hermite spline (cHs) approach. Parameter optimisation in VMStools gave an fm parameter 363 of 0.19, which suggested that a non-linear interpolation gave a more appropriate interpolation of 364 the VMS tracks than a SL interpolation, based on the distance between the interpolated VMS 365 positions and the higher frequency AIS positions. 366 Three days of data were selected at random to display individual interpolated tracks (Figure 5). 367 From visual observation of the three different types of track interpolations (Figure 5), when using 368 a cHs interpolation, the AIS fishing tracks display shorter sections of fishing activity than with the

369 VMS fishing tracks. The low temporal resolution of the VMS data (2 hours) may have forced the

370 interpolations to be continuous such that they potentially missed sections of time in which fishing

did not occur. In contrast, as the AIS data has a higher temporal resolution (5 minutes) it can

account for shorter periods of fishing and steaming within this 2 hour window.

373

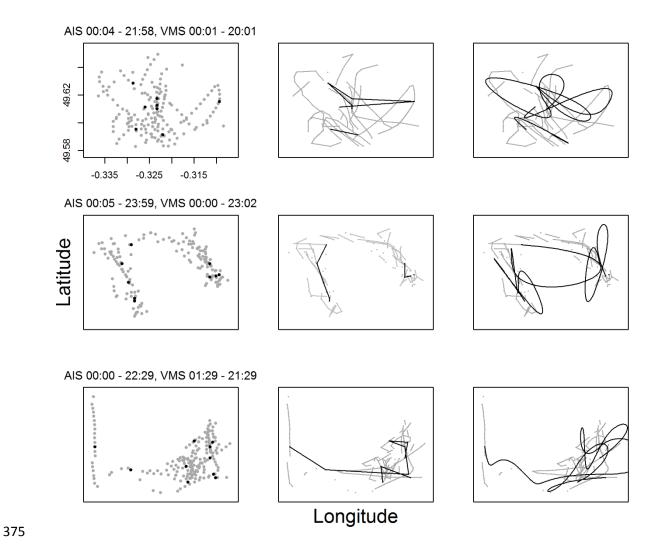


Figure 5. Each row = one trip by one vessel. The time span of each data set is displayed for each row. Black = VMS, grey = AIS. The first column displays the point data, the second column displays straight line (SL) interpolated data, and the third column displays the cubic Hermite spline (cHs) interpolated data for the VMS data and straight line interpolated for the AIS data. The extent of fishing activity appears to be underestimated using a straight line interpolation, but they show a similar extent with a cHs interpolation, albeit with a lower resolution in the VMS tracks.

384

3.4.2. How do the estimates of fishing extent and intensity differ between data types?

385 To investigate how the spatial scale of analysis (i.e. the grid cell size used to aggregate data) 386 influenced the estimates of fishing extent, points that were classed as fishing (cf. steaming or 387 stationary) were joined to spatial grids of varying cell size. Only matching data were used for this 388 part of the analysis. The estimates of fishing extent presented here are not absolute efforts of 389 scallop dredging activity in the English Channel, as data was only available from 8 vessels, and 390 only subsections of data which are temporally matched between AIS and VMS data were used. 391 The analysis is therefore relative, providing a comparison between the data types, and not 392 absolute estimates of fishing extent and intensity. Increasing the grid cell size for analysis 393 increased the estimated extent of fishing activity (Figure 6, Table 2). The total extent estimates 394 were most similar when using the cubic Hermite spline interpolation method for the VMS data. 395 In this analysis an assumption was made that the straight line interpolation of the AIS data using a 396 grid with 1 x 1km grid cells provided the most accurate estimate of the extent of fishing, as it was 397 the highest resolution data treatment. In this case, using a grid with 10 x 10km grid cells 398 substantially overestimated the extent of fishing. The cHs interpolation of the VMS data at 1km 399 cell resolution gave a very similar value for the extent of fishing as the SL AIS at 1km cell 400 resolution, but the point VMS data at 1km cell resolution greatly underestimated the extent of 401 fishing activity. This suggests that the poll frequency of the VMS data is too low to give an 402 accurate estimate of fishing extent unless either points are either aggregated to a low resolution 403 grid, or if using a high resolution (e.g. 1km resolution) grid the points should be interpolated using 404 a cHs interpolation. The method of data treatment had a substantial impact on the estimated 405 extent of fishing when using a high resolution grid (1km resolution) (Figure 6). At a coarse 406 resolution (10km resolution) the method of VMS data treatment had less impact on the estimate 407 of extent, but the overall extent estimate was significantly higher than when using a 1km grid

408 resolution. The data treatment had little effect on the estimate of extent determined from the409 AIS data (Figure 6).

410 The amount of the study area that was perceived as un-trawled (i.e. swept area ratio (SAR) \approx 0)

411 decreased as the grid size increased (Figure 7). Generally, there was very little difference between

412 the different data types (VMS, AIS) or data treatments (point data, SL interpolation, cHs

413 interpolation). There were slight differences in the areas trawled very lightly (SAR < 0.25) when

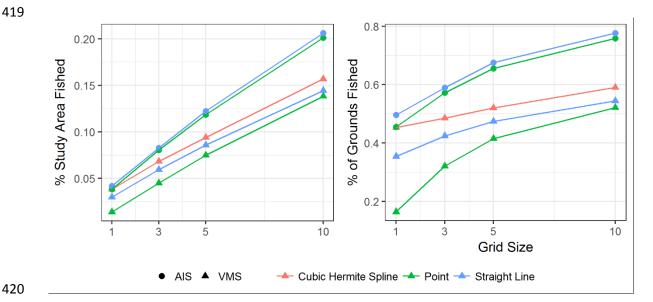
using a high resolution analysis grid (1km cells). Only directly matching data from 7 vessels over

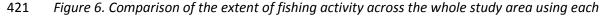
415 12 months was used, therefore although the differences are small, if scaled up to a fleet and

using more complete data, these differences would be multiplied. A 3km grid showed little

417 variation in the pattern of fishing intensity between each data treatment.

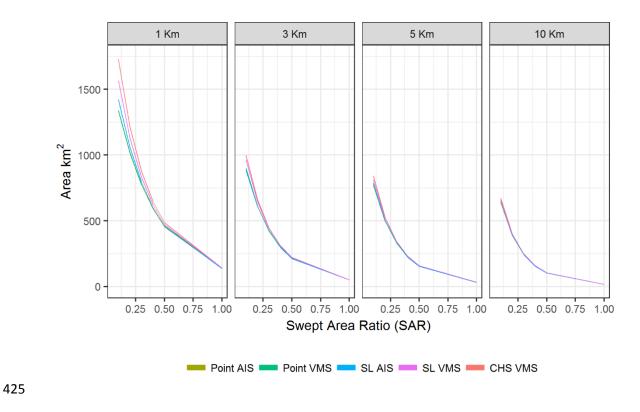






422 interpolation method, and the extent of fishing activity across fishing grounds (i.e. only areas

423 which had recorded fishing activity by any of the data).



426 Figure 7. The area of seabed swept 0 – 1 times during the study period (swept area ratio, SAR), by

427 grid size (1, 3, 5, and 10km), using each method of data interpolation (point, SL and cHs), with only

428 matching VMS and AIS data. Zero SAR data is not included.

Table 2. Comparison of VMS and AIS data using different interpolation methods at different spatial scales. The Difference (km²) column indicates how much larger or smaller the extent was using AIS data, in km², and the percentage difference indicates this difference as a percentage of the study area or the fishing grounds.

		Direct match VMS			Direct Match AIS					
	Grid Size	Extent (km²)	Extent as % of study area	Extent as % of grounds	Extent (km²)	Extent as % of study area	Extent as % of grounds	Differemce (km²)	% difference (study area)	% difference (grounds)
Р	1km	1127	1	16	3123	4	46	1996	2	29
	3km	3654	5	32	6516	8	57	2862	4	25
	5km	6075	8	41	9600	12	66	3525	4	24
	10k	11200	14	52	16300	20	76	5100	6	24
	m									
SL	1km	2432	3	35	3407	4	50	975	1	14
	3km	4833	6	42	6714	8	59	1881	2	17
	5km	6950	9	47	9900	12	68	2950	4	20
	10k	11700	14	54	16700	21	78	5000	6	23
	m									
cHs	1km	3111	4	45				296	0.4	4
	3km	5526	7	49				1188	1	10
	5km	7625	9	52				2275	3	16
	10k	12700	16	59				4000	5	19
	m									

4. Discussion

Neither VMS nor AIS were designed as tools to aid our understanding of fisheries science, but whilst both data sources offer valuable data for understanding fishing activity, their use must be based on an informed understanding of the most appropriate way to process and interpret the data. Of greatest importance, AIS data can have a substantially lower coverage than VMS data and thereby provides a potential underestimate of overall activity (Russo et al., 2016, Natale et al., 2015). This study has contributed to this understanding by demonstrating how even when only looking at VMS and AIS from the same individual vessels (i.e. accounting for the lower fleet coverage of AIS), and even reducing the data to only include fishing trips with concurrent AIS and VMS data logging, there were still substantial gaps in AIS data coverage.

4.1. Can AIS data be an appropriate fisheries monitoring tool?

Whilst AIS data is attracting attention as a promising tool for analysing fishing activity, because it provides publicly available high resolution vessel tracking data, the gaps in its coverage present a substantial hurdle. In this study considerable gaps in the coverage of AIS data compared to the VMS data were found, which concurs with similar studies (Russo et al., 2016, Natale et al., 2015). A similar proportion of VMS and AIS data were retained following data cleaning (i.e. removing incorrect coordinates, points on land, points in harbour etc.), but after only matching data were retained, only 11% of VMS fishing points had corresponding AIS data. This coverage also varied substantially between individual vessels, with 1 – 55% retention of VMS data when using only matching data; for one vessel, there were no directly matching AIS data for 99% of the VMS. For the whole of the studied fleet, this translated as AIS data capturing only 26% of the duration of fishing activity captured by VMS data in 2012. Clearly, this is a substantial gap in the AIS data coverage, and would be a cause for concern if using AIS data to analyse fishing activity without VMS data. It is likely

inappropriate to use AIS data for absolute estimates of fishing extent or intensity, because the gaps in coverage are too substantial. It may be possible to use AIS for comparative studies, making an assumption that the gaps in coverage are relatively consistent over short time frames and on average between fleets. Nevertheless, more comparative studies of AIS and VMS may be needed, to better understand the differences and gaps in coverage. Caution would be required when using historical AIS data compared to more recent or future AIS data, where compliance, increased uptake, or technological advances may potentially lead to higher coverage in recent years.

Aside from addressing the more technical limitations to spatial coverage (Russo et al., 2016; Natale et al., 2015), these results support the suggestions from McCauley et al., (2016) that to gain the full benefits of AIS data for fisheries science, policy interventions would also be required, for example, to reduce the gaps in AIS coverage from fishers turning down the AIS transmitter. There are, however, two principle reasons why fishers may wish to conceal their activity from an AIS system; 1) detection avoidance whilst undertaking illegal fishing activity, and 2) preventing other fishers from using AIS data to identify prime fishing grounds. Real time AIS data is openly available to view, including to other fishermen so it is understandable that fishermen may be reluctant for such high resolution tracking data to be openly and instantaneously available, due to the commercial sensitivity of such data. It is difficult to envisage how this issue would be overcome. If it became a legal requirement that the AIS unit was functioning at full strength and an openly available high resolution high coverage dataset of fishing activity was achieved, it could lead to conflict or negative economic consequences for fishers.

4.2. The Impact of grid cell size and interpolation method

This analysis was designed to highlight any systematic under- or over-estimation of fishing effort due to the low temporal frequency, through comparing it with the higher temporal frequency AIS data.

When VMS and AIS data were directly matching in temporal coverage, the scale of analysis (i.e. size of the grid cells) had the greatest influence on the estimates of fishing extent and intensity compared to the data-type (AIS, VMS) and data treatment (point, SL, cHs). The SAR estimates were very similar, which is perhaps not surprising for temporally matching data over the same time period, unless there were inaccuracies in the GPS positions themselves. Therefore, if the data were truly concurrent, AIS could provide a viable alternative to VMS, and offer the benefit of higher temporal resolution. Nevertheless, the most considerable differences in the data come from the substantial and significant gaps in coverage of AIS data.

When using directly matching data, the VMS data still slightly overestimated the time spent fishing compared to the AIS data, by 129 hours. These differences in recorded fishing duration could be related to the bias from a 2 hourly ping rate of VMS data. Nonetheless, even with directly matching data this constituted 10 extra days of fishing with VMS data (assuming 12 hours continuous fishing per day), or 3% of the total fishing hours recorded. At a small scale of 7 vessels over 1 year of fishing, this is a small value, but scaled up to a whole fleet across multiple years, this could represent substantial fishing effort. The temporal resolution of the VMS data may have missed the shorter hauling/moving sections in between scallop dredge tows, which would be less than two hours in duration, and could therefore overestimate the fishing effort. Identifying fishing activity is much more sensitive under VMS than AIS data, as a 2 hour time frame could be incorrectly classified as fishing or non-fishing, whilst with AIS only seconds or minutes would be misclassified. Technology that would provide information on when the gear is in the water would improve estimates of scallop fishing activity and would address on of the issues of poll frequency for VMS.

4.3. Predicting unpredictable scallop dredging activity

The issues associated with interpolating low poll frequency VMS data are particularly relevant to scallop dredging, due to the idiosyncratic movements of the vessels, such as sharp turns and retowing over the same areas, which may be missed by the lower resolution VMS data (Lambert et al., 2012). The optimal *fm* parameter (a parameter that determines how much the tracks should curve in a cHs interpolation) was markedly different in this study compared to that undertaken by Lambert et al., (2012) with scallop dredgers in the Isle of Man. They concluded that the optimal *fm* parameter was close to zero, i.e. a straight line. Here an optimal fm parameter of 0.19 was reported, which is considerably different from a straight line. For scallop dredgers, and in other fisheries with shorter haul durations, the availability of a higher resolution dataset is perhaps even more important (c.f. trawlers) due to the unpredictable movement patterns associated with these fisheries.

The results presented here therefore relate to scallop dredge fisheries, which represent an atypical part of the European (and global) fishing fleet in terms of towing behaviour characteristics. It may not be appropriate to extrapolate the comparison of matching AIS and VMS data to other mobile bottom contacting fishing gears, such as otter trawlers and beam trawlers. As other trawl vessels perform longer tows, and are characterised by fewer idiosyncratic movements and sharp turns, the discrepancies in estimates from VMS and AIS may be smaller; similar work to compare AIS and VMS data for other types of vessels would enable a better understanding of the value of AIS data for fisheries research.

When providing VMS data in an aggregated format, it is likely not appropriate to provide it at a grid resolution less than 3km by 3km, unless the data has been interpolated, because the extent could be underestimated. It should, however, be noted that the underestimate of fishing extent with point data may be exacerbated by the small sample sizes; where a larger amount of data is available a better estimate of the footprint may be produced without interpolation. Providing data aggregated at a low resolution can overestimate the extent of fishing activity, and interpolation of 2 hourly VMS

pings may be unable to resolve the more complex fishing tracks that some scallop vessels follow. When considering the influence of these factors on fishery management options, there is the need for a balance between data cost (i.e. resolution) and accuracy of the results. Over-estimation of the impacted area may result in more draconian action than is necessary whereas underestimation of the impacted area, whilst of short term benefit to the industry may have longer term repercussions for sustainability.

VMS is often only available in an aggregated format due to confidentiality issues, which can overestimate the extent (Hinz et al., 2013). Highly aggregated VMS data at coarse resolutions is highly limiting in the ability of science to draw reasonable conclusions about fishing footprints and impacts (Hinz *et al.*, 2013). This study has provided a strong argument for the creation of comprehensive positional information at higher temporal resolution than is currently available in order to make robust estimates of fishery activities in space and time. The solution which offers the optimal increase in data accuracy, and therefore accuracy of footprint estimation would be to increase the rate at which VMS data are collected (i.e. higher polling rates). Nevertheless higher quality positional data could also be achieved through more rigorous implementation of AIS units, or through partnership agreements between the scallop industry and scientists. The simultaneous use of AIS and VMS data could also increase the quality of outputs in situations where both data sets are available for analysis (Russo et al., 2016).

5. Conclusions

The present study has highlighted the issues surrounding substantial gaps in coverage by AIS data. McCauley et al., (2016) described AIS as currently a 'service that best observes vessels that don't mind being seen'. This likely arises from a lack of desire to be seen by competing vessels and in some cases may be linked to legal infringements. Nonetheless, VMS already provides a means of collecting

such data in a manner shielded from public view and hence represents an appropriate pathway for the more accurate calculation of fishing footprints through increased polling rates. At present, given the current low frequency of VMS polling there remains the potential for over-reporting of fishing effort, which could lead to a worse assessment of the state of the marine environment in relation to this metric. As AIS was developed for the purpose of safety and collision avoidance, unless additional legislation is effected to regulate the use of AIS by fishing vessels, designed to specifically increase the coverage of fishing activity, it seems unlikely that AIS data could be considered as an equal alternative to VMS data. If the gaps in coverage were addressed, the increased poll frequency of AIS data would allow more accurate analysis of fishing activity, but increasing the poll frequency of VMS data may be a more viable option. The use of reliable high resolution AIS or VMS data would ultimately benefit fishers and scientists, through generating more accurate fishing footprints and a better understanding of the ecosystem impacts of fishing, and thus more sustainable management.

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