Stable isotope-based location in a shelf sea setting: accuracy and precision are comparable to light-based location methods

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Abstract :

1. Retrospective determination of location for marine animals would facilitate investigations of migration, connectivity and food provenance. Predictable spatial variations in carbon and nitrogen isotopes in primary production across shelf seas provide a basis for stable isotope-based location.

2. Here, we assess the accuracy and precision that can be obtained through dietary-isotope-based location methods. We build isoscapes from jellyfish tissues and use these to assign scallops of fixed and known individual location, and herring with well-understood population-level distributions in the North Sea.

3. Accuracy and precision for retrospective isotope-based location in the North Sea were of a similar order to light-based location devices, with 75% of individual scallops assigned correctly to areas representing c. 30% of the North Sea, with a mean linear error on the order of 10(2)km. Applying assignment methods to an alternative migratory species (herring) resulted in ecologically realistic assignments consistent with fisheries survey data.

4. Location methods based on dietary isotopes such as carbon and nitrogen recover the spatial origin of nutrients assimilated into tissues, and this may not correspond directly to the physical location if either the test animal or its prey is highly migratory. Stable isotope-based location can be applied to any marine-feeding organism or derived food product, but the ecological meaning of any assigned area will be more difficult to interpret for large, high trophic level, migratory animals with relatively slow isotopic assimilation rates.

Keywords : assignment, connectivity, geolocation, isoscape, marine, migration, provenance, spatial

44 Introduction

Understanding animal movements is fundamental to population dynamics, 45 46 predator-prey relationships, nutrient and energy fluxes within food webs and 47 management of human-animal interactions. In comparison to terrestrial animals, 48 marine (and aerial) animals encounter relatively few static, physical barriers to 49 movement and dispersal over areas large in comparison to body sizes is a 50 common phenomenon. In the context of marine fisheries, mislabelling of fishery 51 products has emerged as a major problem on global markets (Marko et al. 2004; 52 Wong and Hanner, 2008; Nielsen et al. 2012, Cawthorn et al. 2012). Tracing 53 marine food from origin to sale is a key aim of regulatory organisations 54 worldwide. At present there are few effective retrospective analytical methods 55 available to test claims of spatial origin of traded seafood. 56 Marine spatial ecology is undergoing a revolution with rapid 57 developments in telemetry and electronic tagging technology with the 58 deployment of large static acoustic arrays, satellite geo-location and the 59 development of ever smaller less invasive data storage tags (Hunter et al. 2003; 60 Righton et al. 2007; Block et al. 2011). Nonetheless, direct tagging of marine 61 animals still requires capture and recovery of tags, and processing of data, and is 62 relatively costly (Ramos & Gonzalez-Solis 2012). Furthermore, while tagging 63 experiments reveal individual movements in high resolution, by definition, they 64 cannot be applied retrospectively. Natural tags provide an attractive supplement 65 to direct location tools. Natural location methods typically attempt to link the 66 chemical (or parasite) composition of the test animal's tissues to known spatial 67 variations in the environment (Hobson 1999, Graham et al. 2010, Seminoff et al.

68 2012, McMahon et al. 2013). In recent years, stable isotope location has proven 69 effective at reconstructing long-distance migrations in terrestrial, particularly 70 avian, ecology (Rubenstein & Hobson 2004; Wunder and Norris 2008; Hobson et 71 al. 2012). Statistical models of spatial variation in the isotopic composition of 72 precipitation (Bowen, 2010), vegetation (West *et al.* 2007; Still & Powell 2010,) 73 and higher taxa tissue (Vander Zanden et al. 2015) have been developed in many 74 environments and termed isoscapes. Such isoscapes can provide a base model to 75 assign geographic origin to a tissue of interest, following calibration between the 76 media used to construct the isoscape and the species and tissue to be assigned 77 (Wunder & Norris 2008). A relatively mature literature has developed describing 78 the construction of isoscapes, the statistical considerations surrounding 79 geographic assignment based on isoscapes, and application of isoscapes to track 80 animal movements (West et al. 2010).

81 Isotope-based location comprises a geo-statistical spatial model, a 82 calibration between the model and the species and tissue to be assigned, and a 83 probabilistic comparison between model and measured data. Isoscapes derived 84 from the same species and tissues as those that will be assigned in theory 85 provide the most robust method of assignment. However, the practical and 86 financial limitations associated with sampling and analysing tissues of each 87 migratory species across the full potential foraging range are considerable. 88 Therefore the potential for isotope-based geo-location is greatly increased if 89 multiple taxa can be referred to a single isoscape model. The accuracy and 90 precision available for isotope-based location therefore depends on the variance 91 associated with the underlying geostatistical isoscape model, *in situ* variability in 92 the isotopic compositions of both the organism used to construct the model and

93 in the tissues to be assigned, and the uncertainty inherent in linking the isotopic 94 compositions of the tissues to be assigned to the baseline isoscape (i.e. 95 calibration of the isoscape to the tissue of interest, Wunder & Norris 2008). 96 Considerable debate remains around the most effective way to incorporate error 97 and uncertainty into stable isotope-based geographic assignment methods 98 (Wunder & Norris, 2008; Van Wilgenburg et al. 2011; Wunder 2012; Bowen et 99 al. 2014, Vander Zanden et al. 2015). Wunder (2008) provides a thorough review 100 of the assumptions inherent in isotope-based location, focussing on hydrogen 101 and oxygen isotope based geo-assignment specifically in migratory birds. 102

103 Isotope-based location is not as well developed in marine settings and 104 very few robust assessments of the accuracy and precision obtained using 105 isotope based location have been developed in marine settings (Vander Zanden et al. 2015). In marine systems oxygen and hydrogen isotopes are relatively 106 107 spatially constant, so alternative isotope systems are needed to provide spatial 108 information (Trueman et al. 2012). The isotopic composition of carbon and 109 nitrogen in marine primary production is predictably heterogenous over spatial 110 scales ranging from tens to thousands of kilometres (Jennings and Warr, 2003; 111 Somes et al. 2010; McMahon et al. 2013; Radabaugh et al. 2013, Jennings & van 112 der Molen 2015), and is passed through the food chain. Assigning location based 113 on carbon and nitrogen isotopic compositions therefore effectively tracks the 114 spatial origin of primary production fuelling higher trophic level production 115 rather than the direct spatial location of the animal tested. Nonetheless carbon 116 and nitrogen isotopes have been used extensively to track animal movements

117 across marine isotopic gradients (Hobsen & Schell, 1998; Jaegar *et al.* 2010,
118 MacKenzie *et al.* 2012).

119 Marine carbon and nitrogen isoscape models are generated by 120 interpolation from spatially explicit samples (Schell et al. 1998; McMahon et al. 121 2013). Sessile invertebrates such as filter feeding bivalves have often been used 122 to produce spatial isotope models (e.g. Jennings & Warr, 2003). However, the 123 distribution of sessile invertebrates is limited by water depth and substrate type 124 resulting in systematic variance in spatial coverage of reference samples across 125 the study region. Environmental correlates such as water temperature, depth 126 and salinity have been used to predict isotopic compositions in areas with no 127 reference samples (Jennings & Warr 2003; Barnes et al. 2009; MacKenzie et al. 128 2014), but the resulting isoscape models are strongly dependent on the location 129 of the reference samples and the assumption that regression relationships 130 between environmental drivers and isotope values derived in the sampled 131 region are constant throughout the wider study area. The uncertainty associated 132 with any predicted isotope value increases with (a) the error associated with the 133 regression model, (b) the spatial distance from the reference sites and (c) 134 isotopic or environmental differences between conditions at the predicted site 135 and the mean of the combined reference sites. Estimating the spatially varying 136 uncertainty associated with regression-based isoscape models is not trivial 137 (Bowen & Ravenaugh 2003), and has not been attempted for marine isoscapes. 138 An alternative approach lies in selecting pelagic reference organisms that are 139 widely distributed, but may have larger between-individual variance associated 140 with movement or diet ecology. Scyphomedusan jellyfish provide an attractive 141 potential target due to their ubiquitous distributions, rapid growth and short

142 lifespans (MacKenzie et al 2014). While scyphomedusan jellyfish are mobile, 143 movement is relatively passive and isotopic assimilation rates are fast. The 144 isotopic half-life for the moon jellyfish Aurelia aurita, for example, is estimated at 145 c.10 days (D'Ambra et al. 2014). The distance travelled by jellyfish during the 146 window of isotopic assimilation is therefore likely to be short compared to the 147 spatial scales of isotopic variance in open waters. Jellyfish may be a poor choice 148 for spatial isotope modelling in coastal areas where isotopic variability occurs at 149 smaller spatial scales.

150 Here we assess the precision and accuracy associated with using spatial 151 gradients in carbon and nitrogen isotopes to assign origin to animal tissues 152 across a relatively large shelf sea area. We derive carbon and nitrogen isoscapes 153 from lion's mane jellyfish *Cyanea capillata* expanding on the dataset and 154 methods outlined in (MacKenzie *et al.* 2014, Fig. 1). The North Sea is a shallow 155 semi-restricted shelf sea in the North Atlantic ocean with a total area of around 156 650,000km², sustaining one of the productive fisheries in the world. The North 157 Sea comprises a seasonally-stratified northern basin with a mean depth >50m, 158 and a shallower southern basin that is not stratified. In this study we quantify 159 the accuracy and precision associated with isotope-based geo-location in the 160 North Sea using two independently-determined datasets of stable isotope 161 compositions of the sessile queen scallop *Aequipecten opercularis* (Jennings *et al.* 162 2002, Jennings & van der Molen, 2015). We then identify feeding locations of 351 163 herring *Clupea harengus* caught at known locations throughout the North Sea. 164

165 Materials and methods

167 STABLE ISOTOPE SAMPLES

168

169 Following methods described in MacKenzie et al. (2014), 66 individuals of C. 170 *capillata* were sampled from 52 stations in the North Sea in August 2015 during 171 the International Bottom Trawl Survey on board the RV Cefas Endeavor. Jellyfish 172 were collected, weighed and measured, and a section of bell tissue (mesoglea) 173 removed and immediately frozen. Jellyfish ranged in size from 80 to 240mm in 174 diameter (mean = 107mm, σ = 3.25mm). In the laboratory, tissues were washed 175 3 times with water to remove any soluble nitrogenous materials, re-frozen prior 176 to freeze-drying, sub-sampling and submission for isotopic analyses. Capture 177 locations, body sizes and isotope data for jellyfish tissues are reported in Table 178 S1 and locations are illustrated in Fig. 1.

179

180 351 individual herring were captured at 41 known locations within the North 181 Sea during September 2011 as part of the International Bottom Trawling Survey. 182 Fishing was conducted from the R.V. "Cefas Endeavor". After capture, herring 183 were weighed, dorsal muscle was excised and frozen prior to analysis. Herring 184 under 200mm standard length were grouped as 'small' fish, likely to represent 185 juveniles, whereas fish greater than 200mm standard length are likely to be 186 mature (ICES, 2012). Muscle samples were freeze dried, ground to a powder and 187 analysed for carbon and nitrogen isotopic composition. Capture locations, body 188 sizes and isotope data for herring muscle are reported in Table S2 and locations 189 are illustrated in Fig. 2.

191 Analyses were performed by either OEA laboratories or Elemtex laboratories, 192 Cornwall, UK. Accuracy and precision were monitored through laboratory 193 internal standards (USGS 40 and USGS 41 and a bovine liver standard) and 194 repeat blind analyses of an in-house comparison standard (ARCOS glutamic acid) 195 nested within samples. Accuracy in both laboratories for δ^{13} C and δ^{15} N values 196 was within 0.1‰ of long-term average values for this standard, and precision 197 was 0.2‰ for δ^{13} C and 0.17‰ δ^{15} N values.

198 Jellyfish bell tissue δ^{13} C values showed a significant negative linear 199 relationship with C:N ratios (p = 4.54e-05, slope = -0.047, Adjusted R² = 0.2), 200 implying a variance component related to the concentration of isotopically light 201 lipids within the sample. To correct for potential lipid-related variance in δ^{13} C 202 values, measured δ^{13} C values were adjusted to those predicted for a lipid-free 203 protein (atomic C:N ratio of 3.4) using linear regression between δ^{13} C values and 204 C:N ratios. We did not apply alternative arithmetic lipid correction terms as the 205 measured C:N ratios are close to those expected from pure protein with a small 206 range (mean = 3.6, σ = 0.15) implying that linear corrections are equally 207 effective (Kiljunen *et al.* 2006), and we therefore prefer to use correction terms 208 derived from the species and individuals studied. Lipid-corrected δ^{13} C values of 209 jellyfish show a positive correlation with bell diameter, accordingly they were 210 normalised to the median diameter (107mm):

212
$$\delta^{13}C_{s.cor} = \delta^{13}C_{cor} + (Diameter - 10.73)*0.19$$
 eqn 1
213

Herring muscle contained varying C:N ratios, and δ^{13} C values were corrected for lipid content arithmetically (Kiljunen et al. 2006).

216

217 Isotopic data from queen scallops was recovered from Jennings & Warr 218 (2003) and Barnes et al. (2009) for scallops sampled between 25 July and 29 219 September 2001, and from S. Jennings (pers. comm. 2016) for scallops sampled in 220 similar locations in the summer of 2010 (Jennings & van der Molen, 2015). Up to 221 seven individual scallops were sampled in each area. Locations of capture sites 222 are shown in Fig. 2, Details of sampling, preparation and analytical 223 methodologies are provided in Jennings & Warr (2003), Barnes et al. (2009) and 224 Jennings & van der Molen, 2015. 225 226 STATISTICAL ASSIGNMENT METHODS 227 Statistical models of spatial variation in the isotopic composition of carbon and 228 229 nitrogen in jellyfish tissues sampled in 2015 were drawn from the lipid and size-230 corrected isotope data using Linear Kriging. Isoscapes are presented in Fig. 1 231 together with the associated spatial variances, and locations of jellyfish sampled 232 to create the isoscapes. Raster files of the isoscape values are provided as 233 supplementary data. 234 235 In isotope-based Geo-assignment, the likelihood or probability of the sample 236 originating from a given location or cell in the isoscape depends on the isotopic 237 difference between the sample and cell value relative to the total variance in the

238 isoscape. As described above, much of the difficulty associated with isotope-

239 based location lies in quantifying sources of variance, a problem that is

particularly acute when using environmental correlates to extend predictionsinto regions with no reference samples.

242 As our isoscape model does not contain values predicted from regression 243 models, variance associated with the isoscape is composed of a spatially varying 244 term related only to the physical distance between sample points estimated from 245 the kriging process, and a fixed term reflecting measurement error and between-246 individual variance (Bowen et al. 2014). Measurement error associated with 247 jellyfish analyses determined as the standard deviation from 13 replicate 248 analyses of the glutamic acid standard was 0.2‰ for δ^{15} N and 0.1‰ for δ^{13} C 249 analyses. Between-individual variances in jellyfish isotope compositions were 250 estimated from jellyfish sampled both in 2011 (MacKenzie et al. 2014) and in 2015 as 1.69‰ and 1.04‰ for $\delta^{13}C_{cor}$ and $\delta^{15}N$ values respectively. These 251 252 between-individual variance estimates are similar to those provided for 253 gelatinous zooplankton by Nagata et al., (2015) and Fleming et al., (2015), 254 particularly when accounting for the marked effect of size on isotopic variance in 255 the Fleming *et al.* (2015) data. Total uncertainty in the assignment isoscape was 256 given by:

257

258
$$\sigma^2_{iso}(x,y) = \sigma^2_{k,iso}(x,y) + \sigma^2_{m,iso}(x,y) + \sigma^2_{b,iso}(x,y)$$
 eqn 2

259

260 where $\sigma_{iso}^{2}(x,y)$ is the pooled variance associated with the isoscape prediction, 261 $\sigma_{k,iso}^{2}(x,y)$ is the variance associated with the spatial interpolation model, $\sigma_{m,iso}^{2}$ 262 (x,y) is the variance associated with measurement error and $\sigma^2_{\text{bi.iso}}(x,y)$ is the

263 variance associated with *in situ* between-individual variation.

Measurement error associated with $\delta^{15}N$ analyses of scallop tissues was 264 265 <0.2‰, and the mean standard deviation between individual scallops was 0.8‰, similar to between-individual variance in *C. capillata* δ^{15} N values 266 267 (Jennings & Warr 2003, Jennings & van der Molen 2015). We estimate associated 268 measurement precision associated with δ^{13} C values in scallop tissues as 0.2‰, 269 similar to measurement errors associated with δ^{13} C analyses of *C capillata*. 270 Between-individual variance in lipid-corrected δ^{13} C values of scallops across 22 271 stations sampled in 2010 was 0.21% (Jennings pers. comm. 2016). 272 Pooled error associated with the measurement of scallop stable isotope 273 compositions is therefore given by: 274

275
$$\sigma^2_{assign (x,y)} = \sigma^2_{m.assign (x,y)} + \sigma^2_{bi.assgn (x,y)}$$
 eqn 3

276

277 where σ^{2}_{assign} (x,y) is the pooled variance associated with the isoscape prediction, $\sigma^{2}_{\text{m.assign }(x,y)}$ is the variance associated with measurement error and $\sigma^{2}_{\text{bi.assign }(x,y)}$ 278 279 is the variance associated with *in situ* between individual variation. 280 Uncertainties associated with calibration between the isoscape model and the tissue to be assigned were estimated from the combined uncertainty 281 associated with trophic separation and trophic fractionation between jellyfish 282 283 and scallops (e.g Wunder & Norris 2008). Trophic separation between jellyfish 284 and scallops was constrained from known diet preferences. Scallops are filter-285 feeding molluscs sustained primarily on detrital phytoplankton and

286 microzooplankton. Lion's mane jellyfish are opportunistic pelagic predators 287 consuming a range of macro-zooplankton and larval/juvenile fish. The jellyfish 288 sampled in 2015 encompassed a relatively narrow size range from 80 to 240mm 289 bell diameter equivalent to a wet mass of c. 100-500g, and no systematic size-290 related difference in trophic level between sampled individuals is expected 291 (Fleming et al., 2015). Mass balance (Ecopath) modelling of the North Sea 292 community (Mackinson & Daskalov, 2007) estimates scallop and gelatinous 293 zooplankton trophic levels as 2.8 and 3.6 respectively. We therefore estimate the 294 trophic distance between *C. capillata* and *A. opercularis*, as a single trophic level 295 and assign uncertainty to that estimate with standard deviation of 0.25, ensuring 296 that 95% of the estimates of trophic distance between scallops and *C. capillata* 297 fall between 0.5 and 1.5 trophic levels. 298 Isotopic fractionation between tissue and diet (trophic fractionation) is 299 estimated as 3.4‰ for nitrogen and 1‰ for carbon (Vander Zanden &

Rasmussen, 2001) with a standard deviation of 0.5‰ ensuring that 95% of the

 $301 \qquad \text{estimates of isotopic trophic fractionation fall between 2.4 and 4.4\% for}$

nitrogen and between 0 and 2‰ for carbon. We then created 10,000 trophic

303 fractionation and trophic distance values drawn from the distributions described

above and estimated the distribution of isotopic separation values between

305 jellyfish and scallops.

Scallop muscle and jellyfish bell tissue have contrasting biochemical
compositions and therefore have potential for additional isotopic offsets. We do
not know of any studies reporting isotopic discrimination between jellyfish bell
tissue and coexisting muscle while accounting for trophic level. As all scallops
are known to derive from the isoscape area, trophic-level corrected values

311	should lie within the total range of isotopic vales within the isoscape. We			
312	therefore compare trophic level-corrected scallop data to the full range of $\delta^{13}\text{C}$			
313	and $\delta^{15}N$ values contained in the isoscape, and apply the smallest offset term			
314	required to ensure that all measured scallop values lie within the range			
315	described by the isoscape. For scallops we therefore apply an additional tissue-			
316	specific adjustment of +1‰ (σ = 0.5) for $\delta^{15}N$ and +0‰ (σ = 0.5) for $\delta^{13}C$ values.			
317	The final correction also accounts for any systematic under or over-estimation of			
318	trophic differences or isotopic fractionation. The estimated variance associated			
319	with calibration between scallop and jellyfish tissues $\sigma^{2}_{\text{calib}}$ is therefore			
320	composed of the variance in estimated isotopic spacing across the 10,000 draws			
321	and the estimated variance around the remaining tissue calibration offset			
322				
323	$\sigma^{2}_{\text{calib}(x,y)} = \sigma^{2}(TD^{*}TF_{(x,y)}) + \sigma^{2}_{\text{off }(x,y)} \qquad \text{eqn 4}$			
323 324	$\sigma^{2}_{calib(x,y)} = \sigma^{2}(TD^{*}TF_{(x,y)}) + \sigma^{2}_{off(x,y)} $ eqn 4			
323 324 325	$\sigma^{2}_{calib(x,y)} = \sigma^{2}(TD^{*}TF_{(x,y)}) + \sigma^{2}_{off(x,y)} \qquad \text{eqn 4}$ where x and y refer to $\delta^{13}C$ and $\delta^{15}N$ values respectively, TD is the distribution of			
323 324 325 326	$\sigma^{2}_{calib(x,y)} = \sigma^{2}(TD^{*}TF_{(x,y)}) + \sigma^{2}_{off(x,y)} \qquad eqn \ 4$ where x and y refer to $\delta^{13}C$ and $\delta^{15}N$ values respectively, TD is the distribution of trophic difference values, TF is the distribution of isotopic fractionation values			
323 324 325 326 327	$\begin{split} \sigma^2_{calib(x,y)} &= \sigma^2(TD^*TF_{(x,y)}) + \sigma^2_{off\ (x,y)} \qquad \text{eqn 4} \\ \end{split}$ where x and y refer to $\delta^{13}C$ and $\delta^{15}N$ values respectively, TD is the distribution of trophic difference values, TF is the distribution of isotopic fractionation values and σ^2_{off} is the estimated variance associated with the tissue offset.			
323 324 325 326 327 328	$\begin{split} \sigma^2_{\text{calib}(x,y)} &= \sigma^2(\text{TD}^*\text{TF}_{(x,y)}) + \sigma^2_{\text{off}(x,y)} & \text{eqn 4} \end{split}$ where x and y refer to $\delta^{13}\text{C}$ and $\delta^{15}\text{N}$ values respectively, TD is the distribution of trophic difference values, TF is the distribution of isotopic fractionation values and σ^2_{off} is the estimated variance associated with the tissue offset. We apply the method outlined above to quantify variance terms			
323 324 325 326 327 328 329	$\begin{split} &\sigma^2{}_{\text{calib}(x,y)} = \sigma^2(\text{TD}^*\text{TF}_{(x,y)}) + \sigma^2{}_{\text{off}(x,y)} & \text{eqn 4} \\ & \text{where x and y refer to } \delta^{13}\text{C} \text{ and } \delta^{15}\text{N} \text{ values respectively, TD is the distribution of} \\ & \text{trophic difference values, TF is the distribution of isotopic fractionation values} \\ & \text{and } \sigma^2{}_{\text{off}} \text{ is the estimated variance associated with the tissue offset.} \\ & \text{We apply the method outlined above to quantify variance terms} \\ & \text{associated with assigning herring to the same C. capillata-defined isoscape.} \end{split}$			
 323 324 325 326 327 328 329 330 	σ²calib(x,y) = σ²(TD*TF(x,y))+ σ²off (x,y)eqn 4where x and y refer to δ¹³C and δ¹⁵N values respectively, TD is the distribution of trophic difference values, TF is the distribution of isotopic fractionation valuesand σ²off is the estimated variance associated with the tissue offset. We apply the method outlined above to quantify variance termsassociated with assigning herring to the same C. capillata-defined isoscape.Herring are gape-limited zooplankton feeders with a similar diet and trophic			
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of -0.5 with a standard deviation of 0.5. Multiple individuals were sampled in all

- 41 locations, and mean between-individual standard deviations were 0.44% for
- 336 δ^{15} N and 0.39‰ for lipid-corrected δ^{13} C values. Minimum tissue offset values

between herring and jellyfish were estimated as described above as +2%

- 338 (σ =0.5) for δ^{13} C and +0.5‰ (σ 0.5) for δ^{15} N. A summary of assignment
- 339 conditions is provided in Table 1.

340 We follow the assignment approach outlined in Vander Zanden et al.

341 2015:

$$f(x, y | \mathbf{\mu}_i, \mathbf{\Sigma}) = \frac{1}{\left(2\pi\sigma_x \sigma_y \sqrt{1 - \rho^2}\right)}$$
$$\times \exp\left(-\frac{1}{2(1 - \rho^2)} \left[\frac{(x - \mu_x)^2}{\sigma_x^2} + \frac{(y - \mu_y)^2}{\sigma_y^2} + \frac{2\rho(x - \mu_x)(y - \mu_y)}{\sigma_x \sigma_y}\right]\right)$$

eqn 5

343

342

where $f(x,y|\mu_i,\Sigma)$ represents the probability that an individual with adjusted isotopic composition ($\delta^{13}C=x$ and $\delta^{15}N=y$) originates from a given cell (i) within the isoscape with mean isotopic composition equal to the components of vector μ_i , and variance co-variance matrix Σ . ρ is the correlation between $\delta^{13}C$ and $\delta^{15}N$ values throughout the isoscape, σ_x and σ_y are the pooled standard deviations in $\delta^{13}C$ and $\delta^{15}N$ values respectively given by the sum of the variances:

351
$$\sigma_{(x,y)} = \sqrt{\left(\sigma_{iso(x,y)}^2 + \sigma_{assign(x,y)}^2 + \sigma_{off calib(x,y)}^2\right)}$$
 eqn 6

The range in pooled error terms across the isoscape for scallop assignment was 354 $3.5-12.4\%_0$ for δ^{13} C values and $5.5-11.6\%_0$ for δ^{15} N values, approximately three 355 times higher than the pooled error estimates provided by Vander Zanden *et al.* 356 (2015) where no calibration was needed between isoscape and assignment 357 tissue.

358

359 DISPLAYING ASSIGNMENT OUTCOMES

360

361 The outcome of stable-isotope based location can be displayed as continuous 362 surfaces, but it is easier to describe accuracy and precision based on discrete 363 assignments to a probable area defined by a probability threshold (i.e. an area 364 containing all sites with an assignment probability higher than an arbitrarily 365 fixed value). We use odds ratios to set threshold values (Van Wilgenburg et al. 366 2012, Vander Zanden et al. 2015). The odds of an event occurring is given by the 367 probability of the event occurring relative to the probability of that event not 368 occurring (or P/1-P). Thus a likely event has high odds. Here we define the odds 369 *ratio* as the ratio of odds of the outcome occurring compared to the odds of the 370 most likely outcome possible given the available data:

371

372 Odds Ratio = $(P/1-P_i)/(P/1-P_{i,max})$ eqn 7

373

By setting an odds ratio threshold, all cells with probability values greater than
the threshold are defined as cells of likely origin. The reciprocal of the odds ratio
gives the total proportion of data (and thus the total proportional area) expected
within the threshold limit according to the normal probability density function.

For instance, an odds ratio threshold of 2:1 includes all cells representing the
most likely 2⁻¹ = 50% of all data outcomes and defines a region of likely origin
that is 50% of the total isoscape area. The precision of isotope-based assignment
is thus defined by the odds ratio threshold, and the accuracy is given by the
proportion of assigned individuals where the true location is contained within
the assigned area (Vander Zanden *et al.* 2015).

384

385 **Results**

386 ISOSCAPES

387

388 The spatial isotope models (isoscapes) derived from *C. capillata* are shown in 389 Fig. 1 A,B. Broad spatial patterns are similar to those shown in Jennings *et al.* 390 2003; Barnes et al. 2009; MacKenzie et al. 2014 and Jennings & van der Molen 391 2015, indicating consistent and temporally stable spatial isotopic gradients, and 392 isotopic ranges that are conserved between pelagic and benthic feeding 393 organisms. The newly derived isoscapes are drawn from samples with relatively 394 regular spacing across the modelled area, and the variance associated with the 395 new isoscape models is relatively low and constant across the region Fig 1 C,D. 396 397 ASSIGNMENT ACCURACY AND PRECISION

398

399 The accuracy associated with assigning a geographic origin to the two

400 temporally distinct scallop tissue datasets considering uncertainties in

401 calibration terms and between-individual variance is shown in Fig. 3. The

402 assignment method provides better than random accuracy at all odds ratio

403thresholds (Fig. 3). Assignments are >90% accurate when assigning to areas that404on average represent >40% of the total area of the North Sea. Precision is405enhanced at the expense of accuracy: Doubling the assignment precision to areas406encompassing 20% of the total North Sea reduces accuracy to 50%. The mean407linear error between the cell of maximum likelihood and the known location was408226 (σ = 137) km for the 2001 scallop data and 318 (σ = 114) km for the 2010409scallop data.

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411 HERRING ASSIGNMENT

412

413 Herring were assigned to likely feeding areas using the assignment parameters 414 outlined in Table 1. To report pooled results, individual herring areas were 415 grouped according to body size. Following Van Wilgenburg *et al.* (2011), for each 416 fish, cells designated as likely feeding areas were assigned a value of 1 and all 417 other calls assigned a value of 0. Values were then summed for each cell across the total number of individual fish and divided by the total number of fish, 418 419 providing an index of the most frequently assigned cells ranging between 0 and 1 420 (Fig. 4). Irrespective of capture location, larger fish are assigned to feeding areas 421 in the central northern North Sea (Fig. 4A), consistent with summer fishery 422 catches (ICES 2012, Fig. 4C). Smaller (juvenile) herring are assigned to feeding 423 areas in the southern North Sea particularly around the German Bight (Fig. 4B), 424 again consistent with locations of juvenile herring inferred from acoustic surveys 425 (ICES 2012, Fig. 4D). 426

420

427 **Discussion**

429 ISOTOPE-BASED LOCATION ACCURACY AND PRECISION

430

431	Despite combined uncertainties associated with measurement, between
432	individual variance, and calibration between an isoscape and measured tissues,
433	isotope-based location was 75% accurate to 30% of the North Sea, equivalent to
434	a spatial precision on the order of $10^5~{ m km^{2}}$. The mean linear error between the
435	single cell of highest probability and the known location was between 200 and
436	300 km. Light-based location is widely used in animal ecology, but relatively few
437	studies have tested accuracy of light based location. Where direct tests have been
438	reported, mean errors of location by light range between around 200-400km
439	(Phillips et al. 2004; Lisovski et al. 2012), approximately equivalent to linear
440	errors reported here for isotope-based location methods.

441

The isoscape used here is derived from a mobile pelagic organism, but used to assign origin to a sessile benthic organism collected either 4 or 14 years prior to the samples used to derive the isoscape. This mismatch in sample collection time and organism functional group is deliberate, testing the degree to which isoscapes derived from a single reference organism can be used to assign a wide range of taxa over unspecified periods of time.

448

449 Short and long-term temporal variation in isotopic baselines could confound the

450 use of isotopes for geolocation. Scallops have been sampled in 2001 and 2010,

and jellyfish in 2011 (MacKenzie *et al.*, 2014) and 2015. The regional distribution

452 of isotope values was consistent across these four independent sampling dates,

453 although the exact location of boundaries between isotopically distinct regions 454 varies slightly between sample suites. Consequently assignment accuracy is 455 relatively consistent between the two test datasets (Fig. 3). This is consistent 456 with broad hydrological control over spatial distribution of isotope values, 457 modified by relatively minor intra-year variability (MacKenzie et al., 2014; 458 Jennings & van der Molen 2015). While jellyfish mesoglea sample spring and 459 summer production, scallops have a longer isotopic turnover times and likely 460 integrate annual average production (Jennings & van der Molen 2015). The 461 similarity between jellyfish and scallop isoscapes further supports the argument 462 that spring and summer primary production dominates biomass-weighted 463 consumer tissue production in this strongly seasonal sea. At higher spatial 464 resolution, or in coastal settings, isotopic compositions of primary production 465 and dissolved organic matter are expected to vary more widely in both time and 466 space (Kürten et al. 2013), and the spatial isotope models presented here are 467 unlikely to perform well.

468

469 GEOGRAPHIC ASSIGNMENT OF MIGRATORY FISHES

470

Herring present a particular challenge for fishery management, as they exhibit
complex migratory behaviour and variation in spawning strategies which change
in response to environmental conditions, population sizes, age structures and
harvesting (Dickey-Collas et al. 2010). North Sea herring feed in open waters in
the northern North Sea in summer months, before migrating south and east to
spawn in discrete locations dictated by the need for well-oxygenated coarse
substrates. Larval herring drift eastwards within the southern North Sea

towards the German Bight before recruiting to the adult population. Isotope-

479 based geo-assignment captures this ontogenetic migration (Fig. 4), implying that

480 isotope based location offers a promising additional tool for marine spatial

481 ecology and management.

482

483 IMPLICATIONS FOR ECOLOGY, MANAGEMENT AND FOOD TRACEABILITY484

485 Stable isotope-based retrospective location is well-established in terrestrial 486 ecology, particularly for birds, but extension into marine environments has been 487 slow due to the difficulty of obtaining baseline spatial isotope data. Here we 488 show that isotopic baselines derived from carbon and nitrogen isotopic 489 compositions of pelagic gelatinous zooplankton provide sufficient spatial 490 resolution to rival light-based location in terms of accuracy and precision. 491 Determining location based on carbon and nitrogen isotope compositions 492 records a fundamentally different ecological variable to other location methods. 493 While data storage tag, satellite and water chemistry-based locations record the 494 physical position of the animal, dietary isotope based locations record the likely 495 spatial origin of nutrients assimilated during feeding. In sessile animals, or 496 animals with a limited foraging range, feeding location and physical location will 497 be effectively the same within the error of the assignment methods. In mobile 498 animals (or animals feeding on mobile prey), however, assigned feeding location 499 reflects the origin of primary production assimilated during feeding. Potentially, 500 the location associated with assimilation of food may not necessarily correspond 501 to the location where an animal spends the majority of its time.

502 Dietary isotope-based location provides additional ecological information 503 beyond location at a fixed point in time, but interpreting the ecological meaning 504 of dietary isotope 'location' in migratory animals requires some understanding 505 of the timescale of isotopic assimilation relative to the rate and scale of 506 movements across isotopic gradients. Herring are relatively small, metabolically-507 active, low trophic level (Mackinson & Daskalov 2007) fish, and isotopic 508 equilibration is likely to occur with a half life on the order of c.50 days (Miller 509 2000; vander Zanden et al 2015). Consequently, isotopic-assignment areas for 510 herring closely correspond to feeding areas. Dietary isotope-based identification 511 of feeding grounds will be more problematic in animals where isotopic 512 assimilation rates are slow with respect to movements across isotopic gradients. 513 While static physical location tags (e.g. light or tidal-stream based location) can 514 provide an answer to the question of where animals go (Hammerschlag et al. 515 2011), combinations of physical tags and isotopic location may go some way 516 towards addressing questions of why animals spend time in particular regions. 517 The accuracy and precision of location methods based on carbon and 518 nitrogen stable isotopes is highly dependent on the isotopic calibration between 519 the baseline organism and the species and tissue to be assigned. Estimates of 520 uncertainty associated with all steps in isotopic measurement, spatial modelling 521 and calibration can be quantified and incorporated into assignment algorithms. 522 Calibration methods and uncertainties must be reported with any stable isotope 523 assignment. Nevertheless we suggest that stable isotope based geoassignment 524 can be used in marine systems retrospectively to infer the location where the 525 majority of nutrients were assimilated prior to capture. The method can in 526 theory be applied to any marine feeding organism, but the ecological meaning of

- 527 any assigned area will be more difficult to interpret for high trophic level and
- 528 migratory animals with relatively slow isotopic assimilation rates.

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541

542 Data accessibility

543 Jellyfish and herring isotope data and isoscape raster files are provided in the 544 supplementary information. Stable isotope data from jellyfish, herring and 545 previously published stable isotope data from Oueen scallops sampled in 2001 546 are also available from (doi:10.5061/dryad.609hp). Stable isotope data from queen scallops sampled in 2010 are available from Cefas. The owner of the data, 547 548 Simon Jennings, will be archiving it in Dryad shortly. The final version of this 549 manuscript will include a direct link to this data. If you are viewing the 'Accepted 550 Articles' version after September 2016, please search 'North Sea stable isotope' 551 with Simon Jennings as the author in Dryad to find the relevant data. 552

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753	Supporting Information
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754	Table S1. Locations of capture, bell diameter, bell weight and weight and stable
755	isotope composition (δ^{13} C and δ^{15} N) for <i>C capillata</i> recovered across the North
756	Sea
757	
758	Table S2. Locations of capture and stable isotope composition (δ^{13} C and δ^{15} N) for
759	C harengus recovered across the North Sea
760	
761	C_raster.gri, C_raster.grd R-compatible raster files of the carbon isoscape model
762	
763	N_raster.gri, N_raster.grd R-compatible raster files of the nitrogen isoscape
764	model
765	
766	CVar_raster.gri, CVar_raster.grd R-compatible raster files of spatial variance in
767	the carbon isoscape model
768	
769	NVar_raster.gri, NVar_raster.grd R-compatible raster files of spatial variance in
770	the nitrogen isoscape model
771	
772	
773	Table 1: Assignment conditions adopted for stable isotope based location of
774	scallops and herring against isoscapes derived from jellyfish tissues.
775	

Variable	Isoscape jellyfish	Scallop calibration	Herring calibration
Measurement error	δ^{13} C: 0.1, δ^{15} N: 0.2:	δ^{13} C: 0.2, δ^{15} N: 0.2:	δ^{13} C: 0.2, δ^{15} N: 0.2:

(σ)	measured	estimated / measured	measured
Between-individual	$δ^{13}$ C: 1.69, $δ^{15}$ N: 1.04:	$ δ^{13}C: 0.2, δ^{15}N: 0.7: $	$δ^{13}$ C: 0.2, $δ^{15}$ N: 0.2:
variance	measured	estimated / measured	measured
Trophic distance	NA	1 (σ = 0.25): literature	-0.2 (σ = 0.25): literature
		estimate	estimate
Isotopic trophic	NA	$ δ^{13}C: 1(σ = 0.5), δ^{15}N: $	$δ^{13}$ C: 1(σ = 0.5), $δ^{15}$ N:
fractionation		3.4(σ = 0.5): literature	3.4(σ = 0.5): literature
		compilation	compilation
Tissue specific	NA	$δ^{13}$ C: +0(σ = 0.25), $δ^{15}$ N:	$δ^{13}$ C: +2(σ = 0.25), $δ^{15}$ N:
fractionation		+1(σ = 0.25): graphical	+0.5(σ = 0.25): graphical
		estimate	estimate
Threshold odds ratio	NA	1.33	1.5



- Fig. 1. Isoscape models (A, B) and associated variances (C,D) for δ^{13} C (A, C) and
- 782 δ^{15} N (B, D) values based on *C. capillata* sampled in September 2011. Sample
- 783 stations indicated with filled circles.



- Fig. 2 Locations of herring (open triangles) and scallop (2001 data: filled circles,
- 786 2010 data: open circles) samples within the North Sea



- Fig. 3 Accuracy and precision of assignment for the combined, 2001 and 2010
- scallop datasets. Precision is defined by the probability threshold and expressed
- as the proportion of data (i.e. cells) considered as likely. Accuracy is assessed as
- the proportion of individual scallops where the threshold area contained the
- 792 known sample location.



- represent the number of individual herring assigned to each grid square. A)
- herring >200mm standard length, B herring <200mm standard length. Open
- circles indicate capture locations. C. Spatial distribution of reported landings of
- adult herring (log₁₀ tonnes) in quarters 2 and 3 of 2011, data from (ICES 2012).
- D. Estimated biomass of immature herring in June-July 2011 from combined
- acoustic surveys (ICES 2012).
- 803
- 804