

1 A fully integrated GIS-based model of particulate waste distribution from marine  
2 fish-cage sites.

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18

## 19 ABSTRACT

20

21 Modern Geographical Information Systems (GIS) offer a powerful modelling  
22 environment capable of handling large databases. They are a very suitable  
23 environment in which to develop a suite of tools designed for environmental  
24 management of aquaculture sites, including carrying capacity prediction, land-water  
25 interactions and multi-site effects. One such tool, presented here, is a fully  
26 integrated and validated particulate fish waste dispersion module which uses mass  
27 balance to estimate waste input and takes account of variable bathymetry and  
28 variable settling velocity for feed and faecal components. The model also incorporates  
29 the effect of cage movement on waste dispersion, the first such model to do so.  
30 When tidal range was low (1.67m), the maximum movement of a 22m diameter

31 circular cage was 10.1m and 7.7m easting and northing respectively. Highest  
32 deposition from particulate fish waste is under the cage and incorporation of cage  
33 movement increased the effective area under a cage by 72%. This reduced peak  
34 deposition measurements by up to 32% and reduced the average modelled feed and  
35 faecal settlement at the cage centre by 23% and 11% respectively. The model was  
36 validated by comparing model predictions with observed deposition measured using  
37 sediment traps during three 2-week field trips at a fish farm on the west coast of  
38 Scotland. The mean ratio of observed to predicted waste deposition at 5 - 25m from  
39 the cage centre ranged from 0.9 to 1.06, whilst under the cage the model over-  
40 predicts deposition (observed/predicted = 2.21). Although far-field data was seen to  
41 be comparable the near-field discrepancies resulted in variable overall accuracy in the  
42 model. The overall accuracy based on August 2001 data was  $\pm 50.9\%$ , on February  
43 2002  $\pm 72.8\%$  and on April 2002  $\pm 50.6\%$ . Summarizing the data resulted in an overall  
44 average predictive accuracy of  $\pm 58.1\%$ .

45

## 46 INTRODUCTION

47

48 The effects of waste deposition from fish farm cages have been well studied, in  
49 particular for temperate species such as Atlantic salmon (Salmo salar). Studies include  
50 changes in sediment chemistry (Gowen and Bradbury, 1987; Weston, 1990; Silvert,  
51 1992; Black et al, 1996 Davies et al, 1996; Findlay and Watling, 1997; Kempf et al,  
52 2002), oxygen availability (Enell and Löf, 1983; Hall et al, 1990) and changes in the  
53 number and diversity of benthic species (Brown et al, 1987; Gowen and Bradbury,  
54 1987; Weston, 1990; Henderson and Ross, 1995; Kempf et al, 2002). The extent to  
55 which the seabed is affected depends on the type and quantity of particulate material  
56 being released from the cage site and the local physical conditions, such as  
57 bathymetry and prevailing water currents, both of which can be incorporated into  
58 dispersion models.

59

60 Particulate waste dispersion models can give a cost-effective method to evaluate  
61 outcomes in site selection and biomass limits in terms of local environmental capacity,  
62 to set quality standards and aid decision-making for environmental regulation and  
63 management, by testing a variety of pre-production scenarios for given environmental  
64 conditions. Across Europe the extent to which such models are used for this purpose  
65 varies widely (Henderson et al, 2001). In Scotland, DEPOMOD (Cromey et al, 2002) is  
66 now widely used for Environmental Impact Assessments and to estimate the likely  
67 seabed deposition of in-feed sea-lice treatments (SEPA, 2001).

68

69 Many deposition models of fish cage waste in use are based on an original concept  
70 presented by Gowen et al (1989), who used simple mass balance calculations to  
71 estimate waste levels and dispersion equations in combination with hydrographic data  
72 to assess the downward and lateral movement of particles. Subsequent dispersion  
73 models include fish growth sub-models to more accurately predict waste quantities  
74 (Silvert, 1992; 1994), bathymetry variation (Hevia et al, 1996), settling velocities for  
75 feed and faecal components (Chen et al, 1999a,b; Cromey et al, 2002) and the use of  
76 GIS technology (Perez et al, 2002). The primary purpose of GIS was for the storage,  
77 analysis and display of geographic data. Modern GIS goes well beyond this, however,  
78 and includes a range of powerful spatial modelling and decision making tools which  
79 can be used on a wide range of applications.

80

81 GIS has been established as an excellent tool for facility site selection (Church, 2002)  
82 using spatial analytical approaches with the overlay of thematic data layers, relating  
83 to land function and use, to form an image or graphical output that identifies  
84 appropriate sites. This technology is now widely used in aquaculture site selection  
85 (Ross, 1998; Nath et al, 2000) and is equally relevant for the siting of a range of  
86 aquaculture products and structures such as fish, bivalves, ponds or cages (Congleton  
87 et al 1999; Arnold et al, 2000).

88

89 This paper extends the modelling work of Perez et al (2002), who used a combination  
90 of spreadsheet and GIS to estimate the distribution of fish farm derived particulate  
91 carbon waste. This paper describes a validated particulate waste distribution model  
92 fully integrated into the GIS software by development of a specific programme  
93 module. Such integration in to a GIS-based package is important because it ensures  
94 there is no data loss when integrating data from various sources and the outputs from  
95 the waste dispersion module become one of a number of layers within an integrated  
96 Coastal Zone Management (ICZM) approach to aquaculture site management. As part  
97 of their fieldwork for model validation Cromeey et al (2002) suggest that cage  
98 movement may have accounted for some of the variation in their sediment trap  
99 collections, although the amount was not quantified. The effects of cage movement  
100 are explored and the model is validated by comparison with data collected in the  
101 field.

102

#### 103 **MODELLING PROCEDURE**

104

105 The dispersion module was developed in the IDRISI32 GIS environment (Clark Labs,  
106 Massachusetts, USA), which has been especially designed to allow user extension of its  
107 capabilities. The required code was developed using DELPHI 3 (Borland Software,  
108 California, USA) and the resulting executable was integrated into the IDRISI32 software  
109 using the IDRISI Application Programming Interface (API). The architecture for the  
110 modelling process is shown in Figure 1, which shows the elements developed within  
111 the model and the links between model components, with the general logic of the  
112 model presented below.

113

114 Data for cage block generation, dispersal parameters and mass balance calculations,  
115 are entered into IDRISI32 via two easy to follow dialogue boxes within a waste  
116 dispersion module. Cages may be either square or circular, as part of a block or  
117 separate within a cage array, with the relative layout identified through distance  
118 measures (in m) between cages in a row, between rows and orientation (in degrees)

119 from north; 3 simple characteristics that may be measured at the site(s) of interest.  
120 The final layout of the cages, shown to scale, can be verified visually before  
121 commencement of the modelling process. Cage movement and hydrographic data are  
122 entered by calling spreadsheet files through the dialogue boxes. Settling velocities  
123 are calculated by comparing the known pellet size of the feed used against known  
124 settling velocity distributions (Chen et al, 1999<sup>b</sup>; Cromey et al, 2002). The initial  
125 input of carbon waste from the fish farm through uneaten food and faecal waste is  
126 calculated using a mass balance. Two methods are used, either from total production  
127 biomass and feed conversion rates, or from known feed input. Both methods take into  
128 account percentage carbon in the feed, estimates for carbon lost as production (i.e.  
129 harvested) and carbon lost through respiration and excretion (after Perez et al, 2002).

130

131 Carbon outputs through feed and faecal wastes were treated independently with the  
132 concentrations in each calculated through mass balance. The total quantity of carbon  
133 in each were divided equally between the number of cages and then sub-divided  
134 between each hydrographic measurement (typically measured every 20 minutes over  
135 15-days using an appropriate current meter). This portion is then referred to as a  
136 "packet" of waste. Each packet is dispersed in 3-dimensions based on water depth  
137 (bathymetry) and time-specific current speed and direction (based on Gowen et al,  
138 1989) and random feed and faecal settling velocity. The settling velocities for feed  
139 and faecal particles, for the particular type of feed being used, are calculated using a  
140 technique that randomly selects a settling velocity for each packet of waste from  
141 within the range "mean  $\pm$  1 SD". The effect of varying seabed bathymetry on waste  
142 distribution is included by extracting water depths from digital Admiralty Charts  
143 covering the 250,000 m<sup>2</sup> modelled area in a 50 x 50 cell grid (each cell = 10 m<sup>2</sup>). Half  
144 the average annual tidal range for the area is added to the water depth in each grid  
145 cell to adjust to mean annual water depth.

146

147 Cage movement is registered by temporarily shifting the position of the cage centre  
148 horizontally in X and Y, relative to the cage starting position, by an amount read from

149 the cage movement data file. Initial spatial input of waste is then randomly defined  
150 within this temporary cage area. Distribution of particles commences at the net  
151 depth, removing the need to correct for differences in water speed inside and outside  
152 the cage (Inoue, 1972), the assumption being that the particulate waste is not subject  
153 to lateral movement within the cages. During the modelling of settlement through  
154 the water column, the waste packet is iteratively dispersed in 1m-depth intervals,  
155 based on water flow and particle settling velocity, and stops when packet and water  
156 depth are equal. The quantity of feed or faeces being modelled at the time is  
157 assigned to this grid cell, before the distribution of the next packet of waste begins.  
158 For the next packet of waste the previous cage position is further shifted by reading  
159 from the next line in the cage movement spreadsheet and so on until the whole cage  
160 movement file is used. Vertical and horizontal resolution of movement in the model is  
161 1m.

162  
163 Values of waste settled within specific grid cells is then interpolated, filtered and  
164 finally corrected using the procedure described by Perez et al (2002), before  
165 generation of the final model outputs. The interpolation process assumes that the  
166 first carbon packet deposits in grid cell  $XY_1$ , followed by the next packet in grid  $XY_2$   
167 and so on, based on the 20 minute intervals between hydrographic measurements. In  
168 reality there is a more even distribution between the two points over time, not just at  
169 the two end-points. After iterations are complete, interpolation is used within the GIS  
170 to smooth the distribution of waste. This results in initial over-estimation of the total  
171 deposited wastes, which is finally corrected by the application of a correction factor  
172 (CF) (equation 1, after Perez et al, 2002) that ensures the total amount of waste in  
173 the raster image is equal to the total generated through the mass balance.

174

$$CF = \frac{\text{Total predicted carbon waste (kg)}}{\text{Waste carbon in the image (kg)}} \quad (1)$$

176

## 177 MATERIALS AND METHODS

178

179 The site used for collection of field data and as a basis for the model data was located  
 180 on the west Coast of Scotland and consisted of 12-off 70m circumference (~ 22m dia.)  
 181 circular cages in a 2 x 6 arrangement. Relative to magnetic North the cages were  
 182 orientated at 80°. Each of the cages had a net depth of ~10m. Distance between the  
 183 cage centres within a row was 40m and distance between rows was 48m.

184

### 185 Hydrographic Measurements

186

187 Two Valeport BFM106 current metres (Valeport, Dartmouth, Devon) were deployed  
 188 <100m from the cage site for a complete spring/neap tidal cycle (15 days) in August  
 189 2001. The sampling period was 60 seconds every 20 minutes. Meters were deployed  
 190 in approximately 26m depth on a u-shaped mooring, 3m below surface at the lowest  
 191 predicted tide during deployment and 3m above the seabed. The overall settlement  
 192 vector for each time point during deployment was calculated by averaging flow and  
 193 direction recorded by surface and seabed current meters at each time point. These  
 194 data were used in the model. Data was saved as a comma delimited (.csv) ASCII file  
 195 (current speed, direction) and imported into the model by being called

196

### 197 Measurement of cage movement

198

199 Movement of a single 22m-diameter Polar Circle cage was measured on 4 occasions in  
 200 2002 (16<sup>th</sup> October, 23<sup>rd</sup> October, 29<sup>th</sup> October and 5<sup>th</sup> November) at the fish farm. A  
 201 Wild TC1010 Total Station theodolite equipped with a Leica electronic distance-  
 202 measuring device (Leica AG, Heerbrugg, Switzerland) was used to take measurements

203 of 2 reflectors, positioned on opposite sides of cage every 20 minutes for 8 hours  
 204 inclusive of feeding periods.

205

206 The measurements composed of a horizontal and vertical angle and slope distance  
 207 from a point of origin on the shore. These data were converted into Eastings (Es) and  
 208 Northings (Ns) values (in metres) using Leica's LISCAD Plus Surveying and Engineering  
 209 Environment Software version 4.0 (Leica AG, Switzerland and LISTech, Boronia,  
 210 Victoria, Australia), which gave a resolution of 0.01m. The first reading each day was  
 211 converted to point (0,0) E and N respectively and each subsequent measurement was  
 212 relative to this origin. Two reflectors were used to confirm that each side of the cage  
 213 moved simultaneously and therefore changes in distance were not caused by rotation  
 214 only. All cages were assumed to move by the same amount. Data were incorporated  
 215 in the model as a comma delimited (.csv) ASCII file.

216

## 217 **Model validation**

218

### 219 *Waste input calculation*

220

221 Feed input to a single but representative cage at the field site was measured to an  
 222 accuracy of  $\pm 0.1 \text{ kg day}^{-1}$  using the feedback mechanism from a CAS adaptive feeding  
 223 system (Akvasmart UK Limited, Inverness). In keeping with other models (e.g. Crome  
 224 et al., 2002; Perez et al., 2002), each of the 12 cages at the site was assumed to have  
 225 the same feed input.

226

227 The carbon content of 10 feed pellets (% dry weight (DW)) was measured in triplicate  
 228 ( $n = 30$  in total) using a Perkin Elmer 2400 SeriesII CHNS/O Autoanalyser with  
 229 integrated AD-4 Autobalance on samples weighing 4 - 6mg. Water content of the feed  
 230 was calculated as the difference in weight after drying at 90 °C for 24 hours, as a  
 231 percentage of the original weight ( $n = 10$  for each feed size), being 5% in all cases.  
 232 Feed settling velocity was based on the relationship developed by Chen et al. (1999b)



233 for standard EWOS diets at 10 °C and salinity 33.0. Faecal settling velocity  
 234 distribution was  $0.032 \pm 0.011 \text{ ms}^{-1}$  (after Cromey et al, 2002)

235

236 The level of feed uneaten by fish and lost directly to the environment was set at 3%  
 237 (after Cromey et al, 2002). It was assumed that 14.3% of the carbon consumed was  
 238 used for growth (Chen, 2000) and 60% was respired/excreted (Gowen et al, 1991).  
 239 The remaining carbon was assumed to be incorporated into faeces.

240

#### 241 *Comparison between observed and predicted sedimented carbon*

242

243 Predicted carbon outputs from the GIS-based model were compared against observed  
 244 sedimentation measured in the field using sediment traps. Each trap had 4 replicate  
 245 tubes, with an individual area of  $0.005\text{m}^2$ , for sediment collection. Hydrographic data  
 246 and mass balance data were as specified above. Sediment trap samples were  
 247 collected from the same positions in August 2001, February 2002 and April 2002, every  
 248 3 days over 15-days on each occasion. Sediment traps were positioned using a  
 249 mooring system, as shown in Figure 2, under the cage and at 5m, 15m and 25m from  
 250 the cage edge, in a direction perpendicular to the main water flow and at a distant (~  
 251 800m) reference station.

252

253 Sediment trap samples from each tube were analyzed for total carbon (as % DW) as  
 254 described for fish feed, multiplied by the total DW of the sample and corrected for  
 255 depositional area to give deposition in  $\text{g C m}^{-2} \text{ 3d}^{-1}$ . The 5 samples collected at each  
 256 sampling occasion were added together to give total carbon levels in  $\text{g C m}^{-2} \text{ 15d}^{-1}$ ,  
 257 which was used for comparison against the modelled output. Analysis and observation  
 258 of samples showed no feed pellets were collected in the sediment traps during  
 259 deployment and it was therefore assumed collected sediments were from faecal and  
 260 “background” suspended material only. Carbon levels found within each trap were  
 261 corrected to account for background deposition, which was collected simultaneously  
 262 from a reference station on the specified dates and calculated as described above.

263 Thus model validation was conducted for faecal material only (after Cromey et al,  
264 2002).

265

266 Comparison between observed deposition and modelled deposition was assessed in  
267 two ways. Firstly, as a factor indicating comparability, calculated as

268

$$269 \quad \text{Factor} = \frac{\text{Observed}}{\text{Predicted}} \quad (2)$$

270

271 This was used for comparison at each sampling station at each time point. Secondly,  
272 overall accuracy of the model combining all data for each time point was calculated as  
273 an absolute value using (Cromey et al, 2002)

274

$$275 \quad \text{Overall accuracy} = \frac{\left( \sum \left( \frac{\text{Observed} - \text{predicted}}{\text{Observed}} * 100 \right) \right)}{n} \quad (3)$$

276

277 Where n = number of observation for all stations measured.

278

## 279 RESULTS

280

### 281 Measurement of cage movement

282

283 Data collected on the 5<sup>th</sup> November 2002 was rejected due to poor light resulting in  
284 less than 8 hours of data being collected. Plus and minus distances between dates  
285 were arbitrary as the position of the measuring device varied slightly between each of  
286 the trial dates and the starting position of the cages was arbitrarily set at (0,0).  
287 Maximal variation occurred on 29<sup>th</sup> October at 10.1m and 7.7m, easting and northing  
288 respectively, being up to half the cage diameter, when tidal range was low (1.67m).  
289 Tidal range on all dates was broadly similar (1.61m and 1.87m on 16<sup>th</sup> and 23<sup>rd</sup>  
290 respectively) but the wind on the 29<sup>th</sup> October was stronger and may account for the

higher movement during this period, although this was not measured. Wind on other days was negligible. Overall the movement of the cages was limited by the layout of the moorings and depended on the state of the tide.

Movement of cages resulted in the effective area of deposition directly under cages being increased by 72%, as shown in Figure 3. The spatial starting position and relative settlement position of waste feed and faecal material within the cage would therefore vary with the rise and fall of the tide. This has not been taken into account in available fish farm waste dispersion models used by environmental regulators at present.

#### **Model operation and outputs**

Data input to the model was achieved using the dialogue boxes as a mixture of raw data entry (cage positions, bathymetry, mass balance data) and spreadsheet files (hydrography and cage movement). After data entry the model run time was approximately 10 minutes. Predicted carbon settlement to the seabed was automatically generated within IDRISI as a raster-image, with added legend and bathymetric contours, both of which can be varied to match the specific requirements. Cages could also be added to the output by simply adding a cage layer.

Mass balance calculations showed 3.84 t of particulate carbon entered the marine environment as waste, 3.06 t as faeces and 0.78 t as uneaten feed. Figure 4 (a) shows the predicted distribution of total carbon waste for a model run that does not incorporate cage movement, where peak deposition occurred under the cages at a rate of  $1.55 \text{ Kg C m}^{-2} \text{ 15-days}^{-1}$ . The inclusion of cage movement within the model resulted in predicted deposition level directly under cages being reduced (Figure 4 (b)) to a peak of  $1.07 \text{ Kg C m}^{-2} \text{ 15-days}^{-1}$ . The higher predicted deposition in cages 11 and 12 resulted from the shallower depth of water present under these cages. There

320 was no change in the overall extent of the predicted footprint between each of the  
321 model runs.

322

323 Table 2 shows the average modelled deposition within an area 7m-diameter area  
324 around the centre of the cage starting position and 4.5m-diameter around positions  
325 equivalent to the location of the sediment traps. This was achieved by applying a  
326 mask over the raster-image in IDRISI, which allow data extraction from only the cells  
327 of interest, and averaging the data from each cell. Given the 1m cell resolution used,  
328 averaging over this number of cells provided a more appropriate measure for  
329 comparison than simply choosing a single cell; and also reflected the extent of the  
330 movement experienced by cages, identified above.

331

332 Cage movement reduced the average modelled feed and faecal settlement at the cage  
333 centre by 23% and 11% respectively. Modelled feed dispersion showed little difference  
334 with and without cage movement at distances greater than 5m from the cage edge,  
335 due to feeds high settling velocity, which results in the majority of these particulates  
336 being deposited under or very near to the cage. The combination of current direction  
337 and cage movement resulted in overall deposition increasing slightly in a NNE  
338 direction, as shown by the shift in the “blue” area in Figure 5 (b). This explains why  
339 the feed component of settlement at 5m distance decreased along the transect (Table  
340 2), which was on the opposite side of the cage in a SSE direction. The modelled faecal  
341 dispersion increased in concentration at the 5m station and results from the lower  
342 settling velocity for faeces, allowing time in the model for the quantity that would  
343 have previously been predicted for deposition under the cage to be spread more  
344 evenly in all directions despite the cage movement (Figure 6).

345

#### 346 **Validation**

347

348 Validation was carried out for the integrated GIS model including incorporation of  
349 cage movement. Table 3 provides a comparison between observed and predicted

350 faecal carbon deposition. Variability in predicted carbon deposition at each sampling  
351 station with time was a reflection of variability in production levels giving different  
352 mass balance calculations.

353

354 Observed deposition of nutrient material was shown to be high under the cage and  
355 reduce with increased distance from the cage edge up to 25m. The deposition model  
356 prediction mirrored this high to low gradient. The 'Factor' (observed/predicted)  
357 (Table 3) gives a comparison models' prediction against the observed deposition. For  
358 the most part, the model predictions were higher than the actual deposition, as  
359 indicated by a Factor greater than 1 at the majority of stations. Model predictions for  
360 deposition directly under the cage were considerably higher than observed faecal  
361 deposition. Model predictions were closer to observed deposition as distance  
362 increased from the cage centre (as indicated by the reduction in the factor towards 1  
363 at the 25m station). Thus the model over-predicts deposition at near-field stations,  
364 with an increase in parity between modelled and observed data at the far-field  
365 stations.

366

367 Although far-field data was seen to be comparable the near-field discrepancies  
368 resulted in variable overall accuracy in the model. The overall accuracy based on  
369 August 2001 data was  $\pm 50.9\%$ , on February 2002  $\pm 72.8\%$  and on April 2002  $\pm 50.6\%$ .  
370 Summarizing the data resulted in an overall average predictive accuracy of  $\pm 58.1\%$ .

371

## 372 DISCUSSION

373

374 The particulate dispersion model presented here was targeted at predicting the  
375 distribution of feed and faecal carbon waste, either annually or over the course of a  
376 full production cycle (18 - 24 months), through a wholly integrated GIS-based model.  
377 The model outputs generated for this study covered 15-days of production  
378 commensurate with both available hydrographic data and sediment trap collections

379 used for validation. Although designed with whole production cycles in mind the  
380 model was sufficiently robust to allow variable data and timescales to be simulated.

381

382 Irrespective of their complexity, computer based models are simplified  
383 representations of the processes, variables and relationships that function in the  
384 natural environment. Since their inception for fish cage culture (Gowen et al, 1989),  
385 particulate waste dispersion models have undergone various transformations as the  
386 influences on where particulate waste is deposited on the seabed have become better  
387 understood and the means of modelling these influences has become available  
388 (Silvert, 1992; 1994; McDonald *et al*, 1996; Hevia et al, 1996; Chen et al, 1999a,b;  
389 Cromey et al, 2002). Variable bathymetry, random settling velocity, random particle  
390 starting position and estimates of waste through mass balance generated by the above  
391 work are all included in this GIS-based model. Further, this study has shown that the  
392 movement of cages has a small but important influence on the deposition of  
393 particulate farm waste.

394

#### 395 Sensitivity of the model to cage movement

396

397 Primary sensitivity analysis for this model has been carried out elsewhere (Brooker,  
398 2002) and shows that of the many key parameters tested four, - the effect of constant  
399 verses variable water depth (bathymetry), constant verses variable settling velocity,  
400 changes in percentage feed wastage and changes in FCR, - will have the most effect  
401 on predicted deposition. The extent of that effect is specifically influenced by site  
402 characteristics, feed characteristics and husbandry practice rather than any  
403 underlying universal principle that holds true for all sites.

404

405 In this study the validity of applying cage movement to dispersion models has clearly  
406 been demonstrated and resulted in a redefined distribution of carbon settlement,  
407 lower predicted peak values and a reduction in the predicted particulate settlement  
408 directly under cages. Thus the inclusion of cage movement in waste dispersion models

409 is an important parameter in determining the extent and magnitude of particulate  
410 settlement, especially close to a fish cage. Inclusion of cage movement into  
411 dispersion models, however, is only appropriate when the model has a spatial scale  
412 that can register the movement, which would exclude models using greater than 5m  
413 spatial resolution (Dudley et al, 2000; Cromey et al, 2002). Conversely, although any  
414 spatial resolution can potentially be used in the GIS model used, here a resolution of  
415 1m allowed the extent of the measured movement to be fully integrated and for the  
416 effect to be measurable through the data and images generated.

417

#### 418 **Validation of predicted dispersion with observed sedimentation**

419

420 Model validation is an important function within model development, assessing  
421 agreement between the predictions from the model with data collected in the field  
422 (GESAMP, 1991), whilst at the same time clarifying the assumptions and functional  
423 relationships. The GIS model provided a realistic measure of actual deposition at the  
424 site, giving an average overall accuracy of  $\pm 58.1\%$ , which compares favourably with  
425 other proprietary models, such as DEPOMOD (Cromey et al, 2002) which has a  
426 published accuracy of  $\pm 23.1\%$  at a site with similar water dynamics. Overall,  
427 predictions and observations were a similar order of magnitude and the degree of  
428 accuracy reflected the variability seen at all stations in sediment trap data collections  
429 over the 6 weeks of sampling (data was not shown). Model predictions followed a  
430 similar pattern to field data, with decreasing deposition at increasing distances from  
431 the cage edge and there was no patchiness in the interpolated raster-image.

432

433 The inclusion of a feed loss element in the GIS model was vital for calculating the  
434 quantity of faecal material produced, via the mass balance calculations. Had zero  
435 feed loss been assumed in the mass balance then faecal loss would have been over-  
436 estimated and this is important where validation occurred against the faecal portion  
437 of the modelled output. DEPOMOD (Cromey et al, 2002), for example, calculates  
438 faeces in a different manner, through water content and digestibility, and 100% of the

439 feed is assumed to be eaten resulting in an over-estimation of predicted faecal  
440 carbon, which was not taken in to account during validation. Within the DEPOMOD  
441 model 100% feed consumption is required, however, because only a single model  
442 output is produced, being either total solids or total carbon. The GIS model therefore  
443 has a distinct advantage because feed and faeces are treated independently and  
444 separate raster-images generated, which allows feed loss to be used in the model  
445 even though validation was for the faecal portion only. Feed loss can therefore  
446 correctly be included in the model and allows for a further validation in the future as  
447 more detailed data on spatial and temporal losses of feed becomes available.

448

449 Validation of modelled faecal deposition only is not uncommon (Cromey et al. 2002)  
450 and was carried out because a very high proportion of the sediment trap collections,  
451 spanning 6 weeks of sampling, contained faecal material only as indicated by the  
452 carbon content (data was not shown), with very low feed identified. The use of  
453 faeces only for validation affects the robustness of the model to a certain extent,  
454 especially near to the cages, but exclusion of feed does not significantly affect  
455 predicted deposition at greater distances from the cage because high settling velocity  
456 results in the majority of feed depositing directly under the cage. It is only under the  
457 cage, therefore, that deposition would be expected to be higher than the model  
458 suggests were feed to be included and the sensitivity of the model affected.

459

460 Feed loss is a transient process within cage culture and infinitely depends upon  
461 physical, biological and feeding characteristics at a farm site. The quality of staff  
462 feeding the fish to satiation, the stress on the fish in any one day, the prevailing  
463 weather conditions, tidal speed through the spring-neap cycle, water quality, water  
464 temperature variation with season and level of parasite infestation will all influence  
465 feed loss over varying temporal scales. The model assumes that feed loss occurs  
466 uniformly across all hydrographic measurements, but in reality feed loss is limited to  
467 feeding periods only. Subsequently there is a difficulty in assuming that the feed  
468 element of any deposition model is an accurate depiction of the actual settlement.



469 The best current estimates, for modelling purposes, assess that 3% direct feed waste is  
470 a reasonable assumption based on digestibility data and current husbandry practice at  
471 farm sites (Cromey et al, 2002) with historic estimates (Cho, 1991; Enell and  
472 Ackerfors, 1994) now outdated. Feed loss, specifically when using current husbandry  
473 practice and new technology, is an area that requires further investigation.

474

475 If it is assumed that no errors were present in field collected data, subsequent  
476 measurement of sediment trap contents and model input data then differences  
477 between predicted and observed sedimentation may have been due to processes that  
478 are not included in the model, such as losses from leaching and post-depositional  
479 movement through saltation (Chen, 1999b) and re-suspension (Cromey et al, 2002).  
480 There is also a reliance on 2 dimensional hydrographic data (current speed and  
481 direction) that takes no account of shear stresses between water layers, such as  
482 before and after slack water, eddies and wind generated movement that adds to  
483 turbulent mixing and affects the dispersion.

484

485 There are also elements that are not currently included in any commercially available  
486 or research models, which requires further work to be carried out. Hydrographic data  
487 is measured within 100m of farm sites to represent current speed and direction  
488 through the farm. There is, however, an acknowledged reduction in current speed  
489 and alterations in direction as a result of the presence of nets (Inoue, 1972; Black,  
490 pers. comm.) and fouling of nets over time. Fish may also play a part in distributing  
491 waste, by having a tendency to swim in circles that creates a vortex, giving rise to  
492 suction of water through the bottom of the net and movement away through the cage  
493 at shallow depths (Beveridge, 2004). Such influences may particularly affect the  
494 dispersion directly under cages, the area where the GIS dispersion model predictions  
495 are least accurate. Henderson et al, (2001) noted that all of these processes would  
496 need to be investigated to provide a comprehensive model, with data tested for  
497 sensitivity within the model. Importantly, increasing the validation accuracy under  
498 certain conditions and at certain sites may limit the general applicability of the model

499 to represent species specific cage culture as a whole, which must remain the ultimate  
500 goal of such a model.

501

## 502 General conclusions

503

504 Modern GIS is a powerful modelling environment with the capability to develop user  
505 defined modules as extensions. This was achieved in this work using DELPHI 3 and the  
506 IDRISI Application Programming Interface (API). This capability provides the  
507 opportunity to develop new applications, which can then be processed within the GIS  
508 framework. The output is a set of raster images from which further graphical or  
509 statistical information can be generated depending upon the requirements of the  
510 particular application. The system can operate at any spatial resolution and the 1m<sup>2</sup>  
511 used in this work is particularly suitable for farm level particulate dispersion modelling  
512 and with the potential to use larger scales in an assessment of complex multi-site  
513 systems.

514

515 The model presented here provides easy data entry and a requirement for smaller  
516 data sets, which IDRISI or other GIS software packages are easily capable of  
517 interpolating. Predictive capability in the model enables a range of applications to be  
518 addressed. It allows this model to be used as part of an Environmental Impact  
519 Assessment decision-making process, in determining whether a site is acceptable for  
520 farming, under the banner of site selection (Perez et al, 2003). It is also able to be  
521 used during production for monitoring and to assess the impact of proposed  
522 increases/decreases in production. Although there is an acknowledged need to more  
523 fully understand the nature of fish farm waste settlement and dispersion, the model  
524 presented generally over-estimates which provides a safety net under precautionary  
525 principles in evaluating new site proposals.

526

527 Although this dispersion model provides the industry with a free-standing tool that can  
528 be tested at the farm scale, it has even greater potential when used as part of a suite

of tools designed for environmental management of aquaculture sites, including aspects such as carrying capacity prediction, land-water interactions and multi-site effects. This is an area of on-going research. Importantly, the GIS framework used as the basis for this model allows the integration of varying spatial scales within the same framework. This will be particularly important in the future development of Coastal Zone Management Plans (CZMP) in which waste dispersion is one sub-model (See Ross, 1998; Nath et al, 2000) within a framework that could ultimately provide a fully integrated sustainable decision support system for aquaculture site selection and future development.

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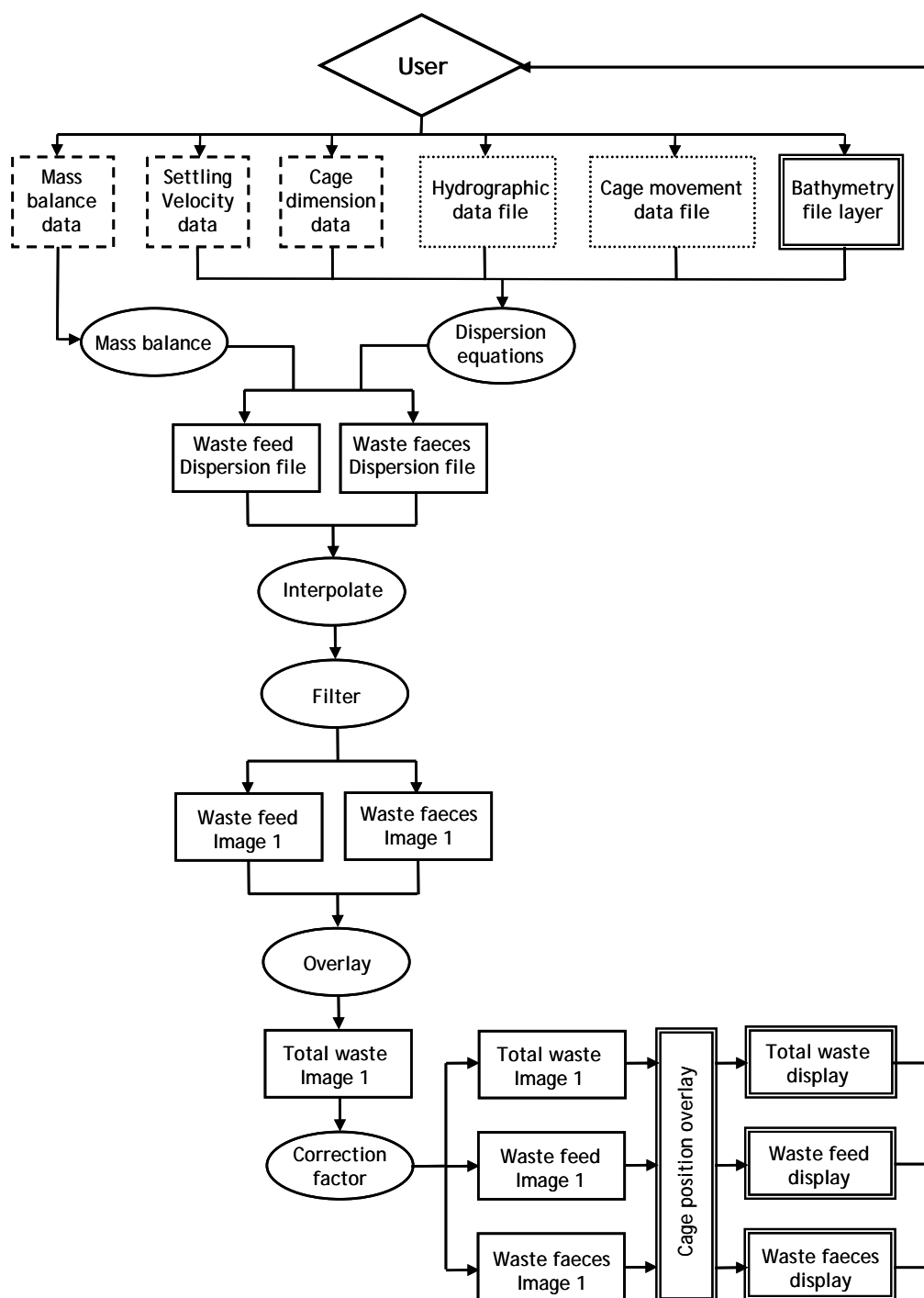


Figure 1: Architecture of the integrated model showing the communication links between the module processes within GIS. Boxes = data, as direct input (-----), as spreadsheet file (.....), as GIS data file ( ——— ) or as a GIS layer ( = ). ○ = GIS process.



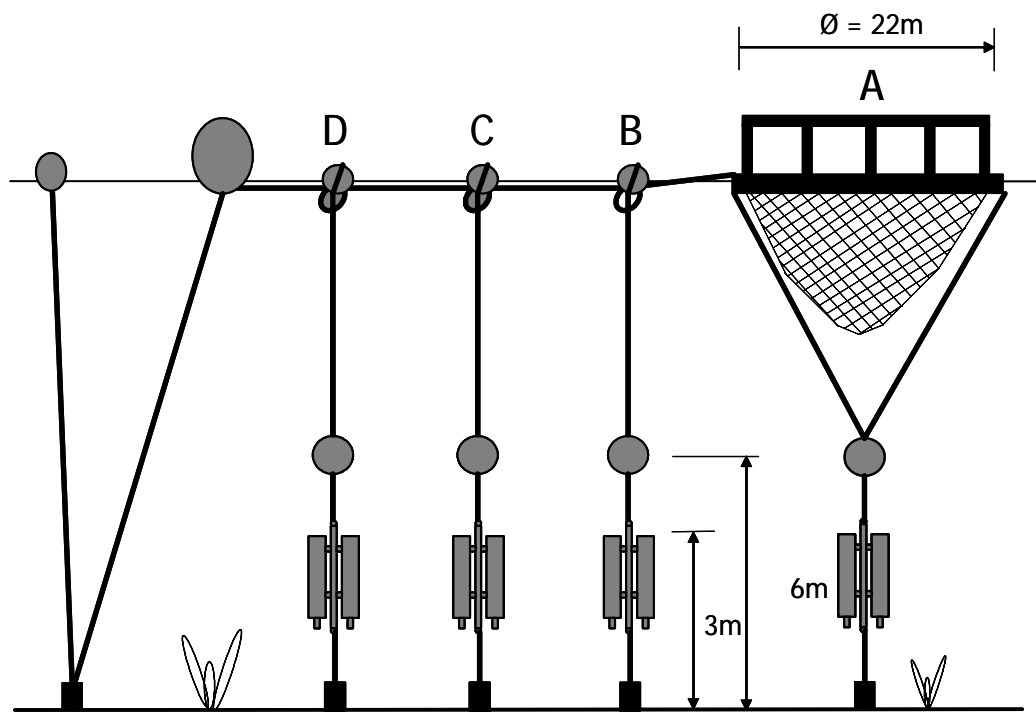


Figure 2: Sediment trap layout on a transect from circular fish farm cage. Traps deployed at distances A = under cage, and B = 5m, C = 15m and D = 25m from cage edge respectively. Not to scale.

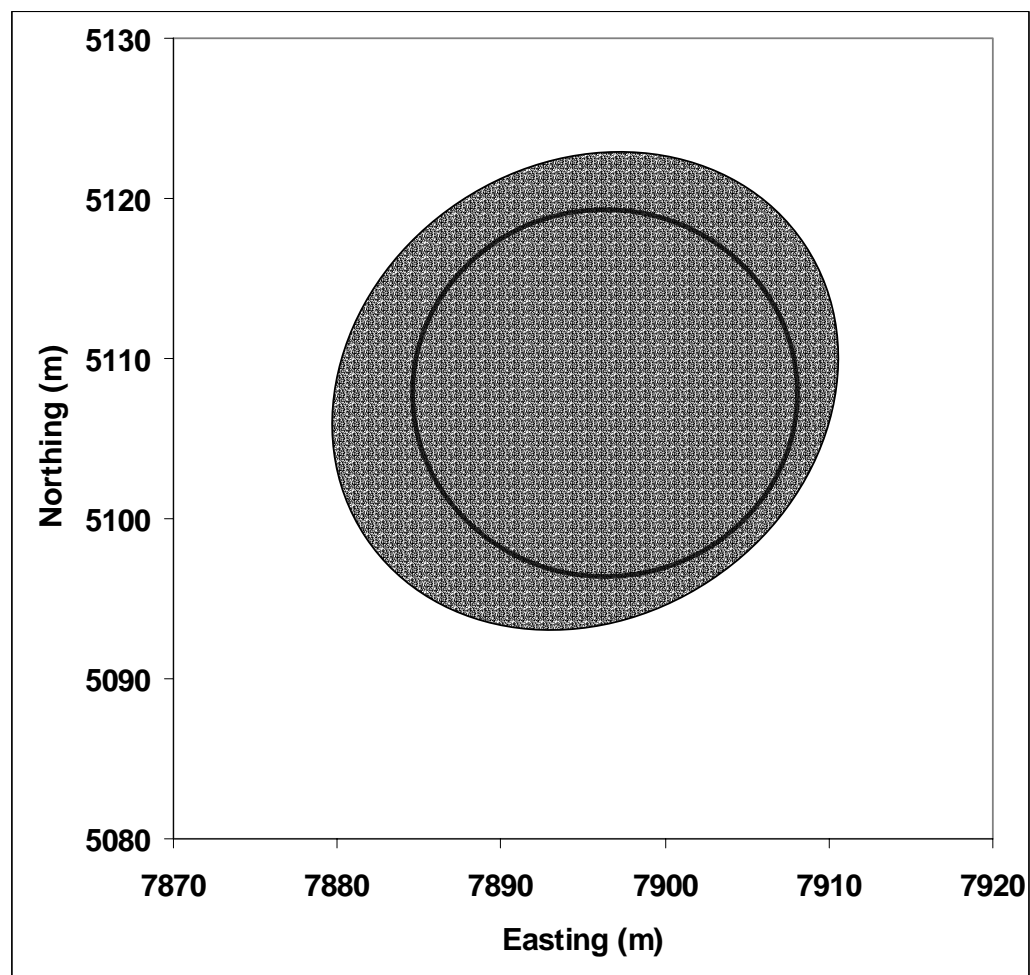


Figure 3: Figure 6.3: Representation of the additional area of seabed covered by a 22m-diameter Polar Circle marine cage as a result of measured movement of the cage on 23<sup>rd</sup> October 2002. Black circle represents cage starting position.

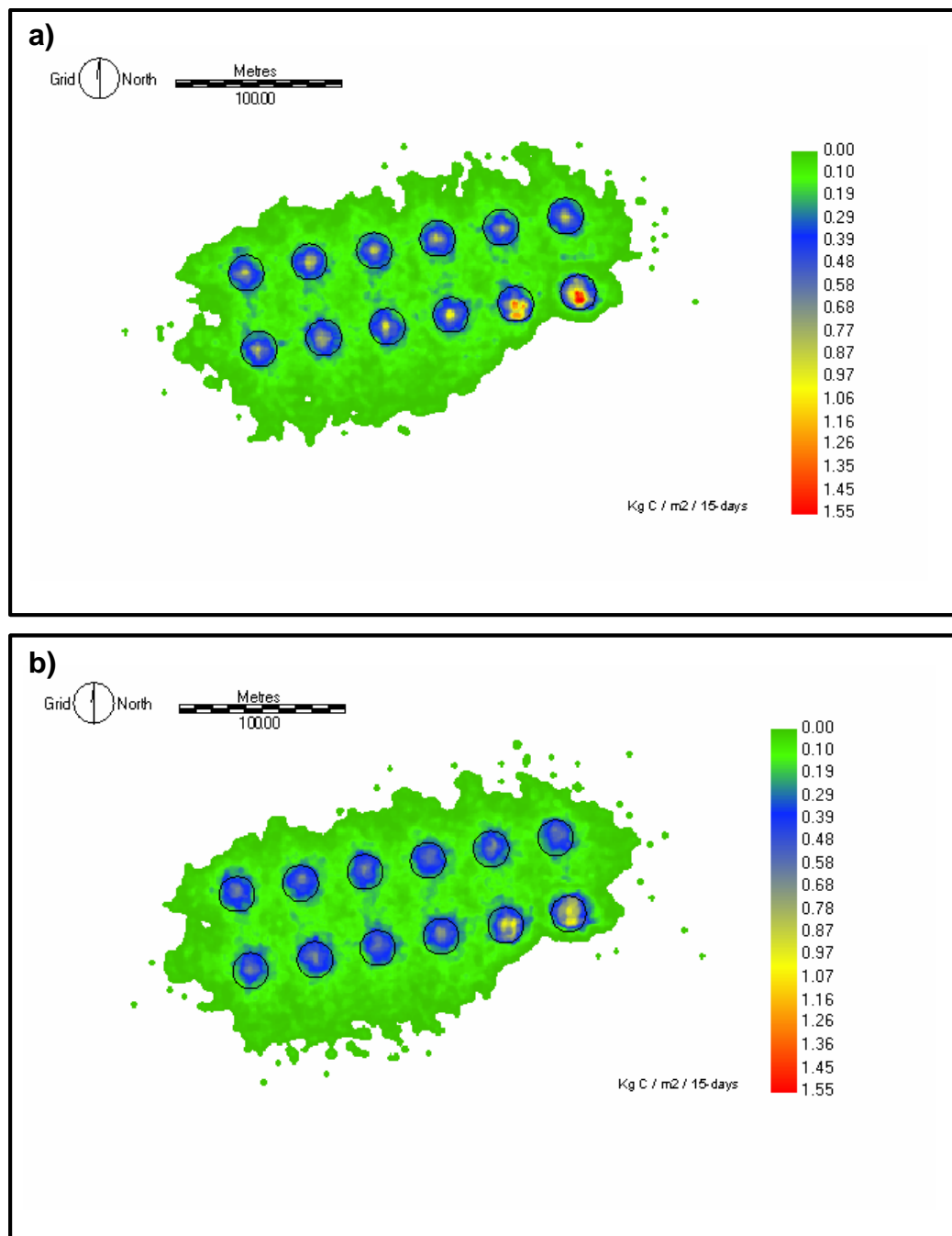


Figure 4: Contour raster-image for fish farm site showing predicted total carbon settlement to the sediment, using GIS dispersion model. (a) static cages model (b) moving cages model.

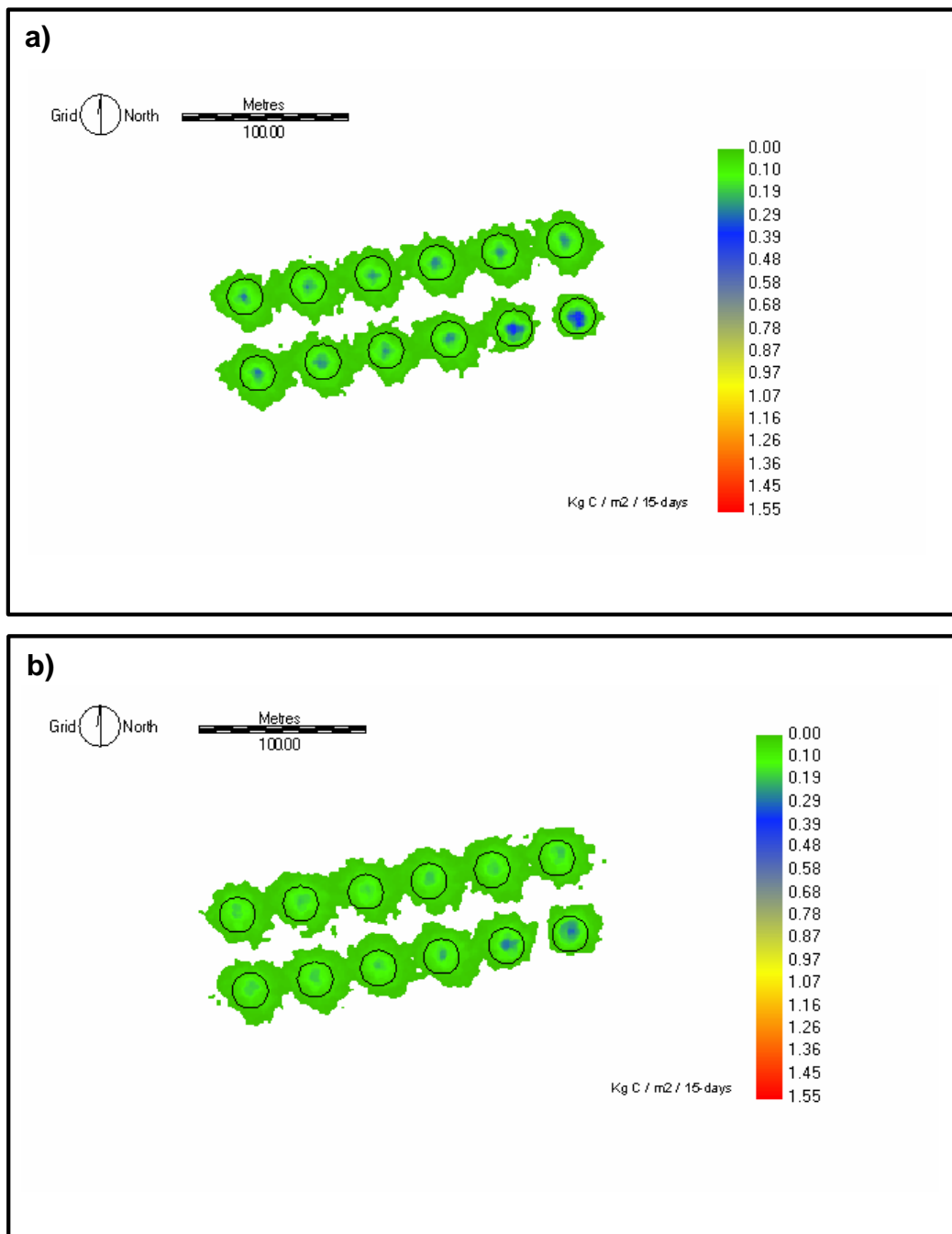


Figure 5: Contour raster-image for fish farm site showing predicted feed carbon settlement to the sediment, using GIS dispersion model. (a) static cages model (b) moving cages model.

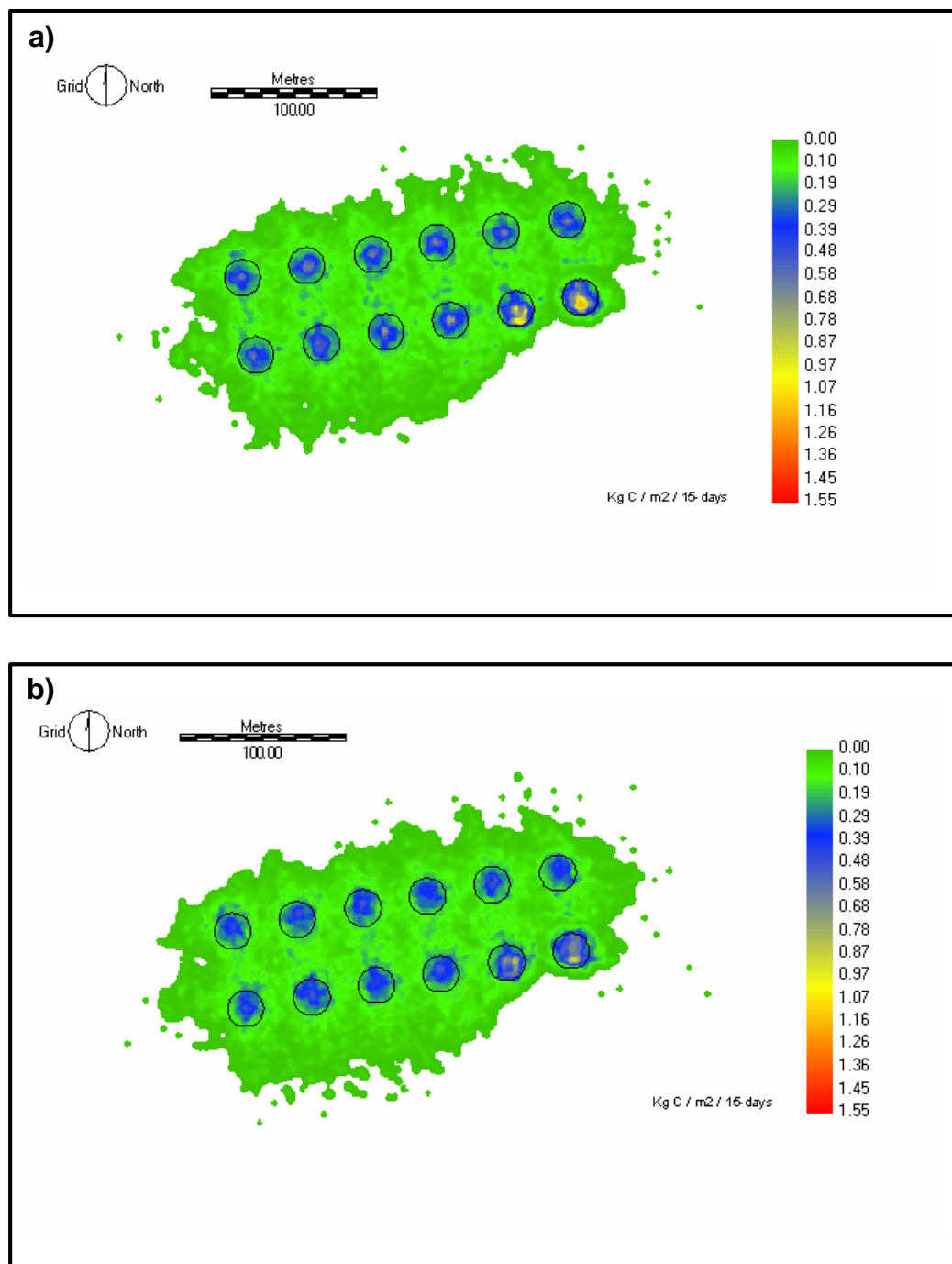


Figure 6: Contour raster-image for fish farm site showing predicted faecal carbon settlement to the sediment, using GIS dispersion model. (a) static cages model (b) moving cages model.

855

856 Table 1: Mass balance data used in waste dispersion model for 15-day trial periods at fish farm site.

857

Trial date	Production in trial cage (kg)	Feed input (kg)	FCR	Feed size (mm)	Mean feed settling velocity (cm s <sup>-1</sup> )	Feed carbon content (% DW)
August 2001	3964	4360	1.10	3 and 6	8.26	51.0
February 2002	2983	3460	1.16	9	10.81	49.5
April 2002	2814	3152	1.12	9 and 12	12.92	51.0

858

859 Production = fish growth between start and end of experimental periods from growth curves and feeding algorithms within a CAS Adaptive Feeding System  
 860 (Aquasmart UK Limited, Inverness).

861

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864

865 Table 2: Average predicted deposition under and at specified distances from the edge of fish cage. Predictions generated using GIS dispersion model  
 866 assuming static and moving cages, based on production and mass balance for the period August 16<sup>th</sup> - 31<sup>st</sup> 2001. Units: g C m<sup>-2</sup> 15-days<sup>-1</sup>.

867

Component	Under cage		5m		15m		25m	
	Static	moving	static	moving	static	moving	static	moving
Faeces	480.71	426.60	115.04	129.04	59.71	58.76	24.01	27.45
Feed	216.81	166.89	38.77	21.81	1.94	1.04	0.23	0.19
Total	679.51	593.50	153.81	150.86	61.65	59.80	24.24	27.65

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Table 3: Comparison of 15-day measured observations verses predicted faecal particulate carbon deposition. Observed deposition measured using sediment traps. Predictions generated using a GIS dispersion model, incorporating cage movement and based on mass balance for 15-days production in tonnes. FCR = Feed Conversion Ratio. Station distance = distance from cage edge (m). Factor = observed/predicted. Number of cells averaged under cage (n) = 38, at remaining stations n = 16. Units: g C m<sup>-2</sup> 15-days<sup>-1</sup>.

Collection	Production (t)	FCR	Under cage			5m station			15m station			25m station		
			Obs.	predicted	Factor	Obs.	predicted	Factor	Obs.	predicted	Factor	Obs.	predicted	Factor
August 2001	3.84	1.10	234.3	426.6	1.82	75.8	129.0	1.70	41.0	58.8	1.43	29.8	27.5	0.92
February 2002	3.06	1.16	85.2	310.7	3.65	120.8	133.8	1.11	55.6	51.0	0.92	22.5	24.3	1.08
April 2002	2.82	1.12	159.6	323.3	2.03	109.5	61.6	0.56	61.7	39.1	0.63	49.5	39.8	0.81
Average			159.7	353.5	2.21	102.0	108.1	1.06	52.8	49.6	0.93	33.9	30.5	0.90