

1 Fukushima  $^{137}\text{Cs}$  releases dispersion modelling over  
2 the Pacific Ocean. Comparisons of models with water,  
3 sediment and biota data

4  
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6 **Abstract**

7 A number of marine radionuclide dispersion models (both Eulerian and La-  
8 grangian) were applied to simulate  $^{137}\text{Cs}$  releases from Fukushima Daiichi nuclear  
9 power plant accident in 2011 over the Pacific at oceanic scale. Simulations extended  
10 over two years and both direct releases into the ocean and deposition of atmospheric  
11 releases on the ocean surface were considered. Dispersion models included an embed-  
12 ded biological uptake model (BUM). Three types of BUMs were used: equilibrium,  
13 dynamic and allometric. Model results were compared with  $^{137}\text{Cs}$  measurements  
14 in water (surface, intermediate and deep layers), sediment and biota (zooplankton,  
15 non-piscivorous and piscivorous fish). A reasonable agreement in model/model and  
16 model/data comparisons was obtained.

17 *Keywords:* Fukushima-Daiichi accident; dispersion model; ocean; sediment; biological  
18 uptake model; caesium

19 **1 Introduction**

20 After the 9.0 magnitude earthquake and resulting tsunami occurred on March 11th, 2011,  
21 in Japan, significant amounts of radioactive material were released to the environment  
22 from Fukushima Dai-ichi nuclear power plant (FDNPP). Radionuclides released to the  
23 atmosphere were transported eastward by a strong jet stream and reached the coast of  
24 North America in four days (Takemura et al., 2011). A portion of these radionuclides was  
25 deposited on the Pacific Ocean surface by wet and dry deposition processes. In addition,

26 water used to cool a damaged nuclear reactor leaked into the ocean (Kobayashi et al.,  
27 2013).

28 Some exercises comparing numerical model performances when applied to simulate the  
29  $^{137}\text{Cs}$  releases from FDNPP in the Pacific Ocean have been carried out, as for instance in  
30 Masumoto et al. (2012). These authors found discrepancies between the five participating  
31 models and concluded that they were due to the different calculated current fields in the  
32 coastal waters of Japan, off Fukushima, which lead to different radionuclide distributions.  
33 Differences in circulation fields were caused by the different ocean models and dispersion  
34 model settings used by the research groups. However, a systematic assessment aimed at  
35 investigating the reasons of differences was not carried out.

36 The Science Council of Japan (SCJ, 2014) carried out a similar intercomparison study  
37 for  $^{137}\text{Cs}$ , with eleven models involved. Again, significant differences between models  
38 were found. Models were different in concept (Eulerian vs. Lagrangian), with different  
39 setting and even different source terms. It was concluded that a simple comparison was  
40 not straightforward and consequently detailed systematic comparison studies, such as  
41 ones that use the same radionuclide forcing with different models and/or the same model  
42 with different forcing scenarios, were required. This kind of intercomparison exercise was  
43 carried out in the frame of IAEA (International Atomic Energy Agency) MODARIA<sup>1</sup> pro-  
44 gram (Periáñez et al., 2015a; 2016a). The MODARIA project was running from 2012 to  
45 2015 to make progress in the assessment of radioactive substances in the environment and  
46 its impact to man and biota. Different dispersion models were applied to simulate FDNPP  
47 releases in the Pacific, using different and also the same water circulation fields. Simu-  
48 lations with the same set of parameters (like diffusion coefficients for instance) were also  
49 carried out. It was found that the main source of discrepancy between different dispersion  
50 models was due to the different circulation fields. Model/model and model/measurements

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<sup>1</sup>Modelling and Data for Radiological Impact Assessments. Further information can be found here:  
<http://www-ns.iaea.org/projects/modaria/default.asp?l=116>

51 comparisons for both the dissolved phase and bed sediments (not included in earlier model  
52 comparison exercises) were carried out in such study.

53 Alternatively, the same dispersion model forced with different circulation fields was  
54 tested as well, although water/sediment interactions were not included in this study  
55 (Kawamura et al., 2017).

56 An interesting exercise was described in Maderich et al. (2018). In this case the  
57 same dispersion model, running with generic parameters, was applied to describe  $^{137}\text{Cs}$   
58 dispersion from Chernobyl NPP accident in the Baltic and Black seas, and FDNPP acci-  
59 dent in the Pacific Ocean. The applied box model (POSEIDON-R; Lepicard et al., 2004;  
60 Maderich et al., 2014a; 2014b; Bezhenar et al., 2016) contained an embedded food web  
61 model. Comparisons of model results with measurements in the three scenarios indicated  
62 that, with some restrictions, the model could be used with generic parameter values in  
63 radiation emergency situations in areas where limited information is available.

64 MODARIA-II program<sup>2</sup> was launched by the IAEA in 2016 as a follow-up of MODARIA.  
65 The work in comparing numerical model performances when applied to simulate FDNPP  
66 releases in the ocean was continued in the frame of this project. Nevertheless, spatial range  
67 and temporal frame of simulations were extended: two year long simulations over almost  
68 the whole North Pacific Ocean were carried out. In addition, marine dispersion models  
69 contain an integrated biological uptake model (BUM) with four components (phytoplank-  
70 ton, zooplankton, non-piscivorous and piscivorous fish). Model/model and model/data  
71 comparisons were carried out for water, bed sediments and biological components of the  
72 models, which has not been done before.

73 Six institutes have participated in the model comparisons. These are the Institute  
74 of Mathematical Machines and System Problem (IMMSP, Ukraine), Korea Institute of  
75 Ocean Science and Technology (KIOST, Rep. of Korea), ABmerit (Slovakia), Univer-  
76 sity of Seville (USEV, Spain), Japan Atomic Energy Agency (JAEA, Japan), and Korea

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<sup>2</sup><http://www-ns.iaea.org/projects/modaria/modaria2.asp?s=8&l=129>

77 Atomic Energy Research Institute (KAERI, Rep. of Korea).

78 The methodology is presented in section 2, where water circulation used by models,  
79 source terms, and the origin of experimental data on  $^{137}\text{Cs}$  concentrations are described.  
80 Results are presented in section 3. Some general discussion on model uncertainty and  
81 complexity is finally included in section 4.

## 82 **2 Methods**

### 83 **2.1 Hydrodynamics**

84 Water circulation provided by FORA<sup>3</sup> model was used for calculations. This model,  
85 Four-dimensional Variational Ocean ReAnalysis for the Western North Pacific (FORA-  
86 WNP30), is the first-ever dataset covering the western North Pacific over the last three  
87 decades (1982-2014) at eddy-resolving resolution. It is a cooperative work of the Japan  
88 Agency for Marine-Earth Science and Technology (JAMSTEC) and the Meteorological  
89 Research Institute, Japan Meteorological Agency (JMA/MRI) using the Earth Simulator  
90 (Usui et al., 2017; Tsujino et al., 2010).

91 The domain used in the present calculations extends  $117^{\circ}\text{E}$ - $160^{\circ}\text{W}$  and  $15^{\circ}\text{N}$ - $65^{\circ}\text{N}$  in  
92 longitude and latitude, respectively. Horizontal resolution is  $0.1^{\circ}$  and there are 54 vertical  
93 levels (0-6300 m) with increasing thickness from the surface to the sea bottom. Monthly  
94 climatological data from 2011 to 2014 were used. Two year long (March 11, 2011 to March  
95 11, 2013) simulations were made.

96 The model domain showing water depths and an example of surface water circulation  
97 (averaged value for March 2011) can be seen in Fig. 1. The general large scale circulation  
98 in the western Pacific Ocean is dominated by the interaction between the Kuroshio and  
99 Oyashio currents. The Kuroshio Current is the western boundary current in the north  
100 Pacific, which flows along the coast of Japan towards the north and curves to the central

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<sup>3</sup><http://synthesis.jamstec.go.jp/FORA/e/index.html>

101 Pacific Ocean, then forming the so-called Kuroshio Extension. The Oyashio Current is  
102 a cold current which flows from the north. These two current systems converge in the  
103 coastal waters off Fukushima coast. Such convergence leads to the generation of unsteady  
104 eddies in the area. These features may be seen in Fig. 1.

## 105 **2.2 Radionuclide sources**

106 Radionuclides were directly introduced into the Pacific Ocean from FDNPP. They were  
107 also released to the atmosphere; radionuclides which were later deposited on the sea  
108 surface. Both sources were considered in calculations.

109 Direct releases of  $^{137}\text{Cs}$  are given for the period March 25th, 2011, to December 31th,  
110 2011, and presented in Fig. 2. They were reconstructed by JAEA as explained in de-  
111 tail in Kobayashi et al. (2013). Monitoring data from the web site of Tokyo Electric  
112 Power Company (TEPCO), regarding the area near the northern and southern discharge  
113 channels of the Fukushima Daiichi NPP (TEPCO, 2011), were used for this purpose.

114 Atmospheric deposition in the North Pacific Ocean was obtained from the averaged  
115 values from WSPEEDI-II (JAEA: Terada et al., 2012) and LADAS (KAERI: Suh et al.,  
116 2006; Suh et al., 2009) atmospheric dispersion models for the period March 12th, 2011,  
117 to June 1st, 2011. Even though simulations are 2 year long, most deposition occurred  
118 within the first months after the accident. Daily integrated values were provided. As an  
119 example, the integrated deposition for March 15th, 2011, averaged from both models, is  
120 presented in Fig. 2.

121 In addition, a pre-FDNPP accident  $^{137}\text{Cs}$  uniform background of  $1.5 \text{ Bq/m}^3$  was con-  
122 sidered over the Pacific Ocean waters, in order to carry out comparisons of model results  
123 with field measurements.

## 124 **2.3 Dispersion models**

125 Some of the main characteristics of the dispersion models which were applied are summa-  
126 rized in Table 1. Both Eulerian and Lagrangian models were used with different param-  
127 eterizations of horizontal and vertical diffusivities. The general characteristics and basic  
128 equations describing the two types of dispersion models which were applied are presented  
129 in appendix A.1 and A.2.

130 A kinetic (dynamic) approach was applied to describe water/sediment interactions in  
131 both Eulerian and Lagrangian models, which is based on a desorption coefficient and the  
132 distribution coefficient,  $k_d$ , of the corresponding radionuclide (Perianez, 2005).

133 All models used an equilibrium distribution coefficient of  $2.0 \text{ m}^3/\text{kg}$ . This is the  
134 mean value recognized by IAEA (2004) for open ocean waters and is also in agreement  
135 with measurements off Fukushima (Honda et al., 2012). The kinetic rate describing  
136 release from sediments,  $k_2 = 1,16 \times 10^{-5} \text{ s}^{-1}$ , was determined for Cs from experiments  
137 (Nyffeler et al., 1984). The kinetic rate describing uptake ( $k_1$ ) is derived from  $k_2$  and the  
138 distribution coefficient, as usually done (Perianez, 2005). A stochastic method is used to  
139 solve uptake/release processes in Lagrangian models (Perianez and Elliott, 2002).

140 Most models include a biological uptake model (BUM). Four species were considered:  
141 phytoplankton, zooplankton, non-piscivorous and piscivorous fish. Three types of BUM  
142 were used in the models: an equilibrium model based upon a concentration factor  $CR$   
143 (appendix B.1), a dynamic model (B.2) and an allometric method (B.3). The BUM  
144 incorporated within each physical dispersion model is indicated in Table 1 as a reference  
145 to the appendix where the corresponding BUM characteristics are commented.

## 146 **2.4 Experimental data**

147 Model results were compared with available  $^{137}\text{Cs}$  measurements in water at three different  
148 layers, bed sediments and biological compartments (zooplankton, non-piscivorous and

Model	I/K THREETOX (IMMSP/KIOST)	I/K Lagrangian (IMMSP/KIOST)	ESTE (ABmerit)	USEV (Univ. Seville)	SEA-GEARN (JAEA)	LORAS (KAERI)
Model type	Eulerian	Lagrangian	Lagrangian	Lagrangian	Lagrangian	Lagrangian
Reference	Maderich et al. (2016)	Brovchenko et al. (2018)	www.abmerit.sk	Periáñez et al. (2016b)	Kobayashi et al. (2007)	Min et al. (2013)
Horizontal diffusion	Smagorinsky <sup>a</sup> formula	10 m <sup>2</sup> /s	Smagorinsky formula	Smagorinsky formula	10 m <sup>2</sup> /s	10 m <sup>2</sup> /s
Vertical diffusion	10 <sup>-4</sup> m <sup>2</sup> /s	10 <sup>-3</sup> m <sup>2</sup> /s for $d < 60$ m 10 <sup>-5</sup> m <sup>2</sup> /s for $d > 120$ m linear function for 60 < $d < 120$ m	10 <sup>-4</sup> m <sup>2</sup> /s	10 <sup>-4</sup> m <sup>2</sup> /s	10 <sup>-4</sup> m <sup>2</sup> /s	10 <sup>-3</sup> m <sup>2</sup> /s
Bed porosity	0.6	0.6	0.6	0.6	0.6	0.7
Sediment thickness	0.05 m	0.05 m	0.05 m	0.05 m	0.05 m	0.1 m
Particle density	2600 kg/m <sup>3</sup>	2600 kg/m <sup>3</sup>	2600 kg/m <sup>3</sup>	2600 kg/m <sup>3</sup>	2600 kg/m <sup>3</sup>	2600 kg/m <sup>3</sup>
<sup>137</sup> Cs $k_d$	2 m <sup>3</sup> /kg	2 m <sup>3</sup> /kg	2 m <sup>3</sup> /kg	2 m <sup>3</sup> /kg	2 m <sup>3</sup> /kg	2 m <sup>3</sup> /kg
$k_2$ (s <sup>-1</sup> )	3.17 × 10 <sup>-8</sup>	Maderich et al. (2017)	1.16 × 10 <sup>-6</sup>	1.16 × 10 <sup>-6</sup>	1.16 × 10 <sup>-6</sup>	1.16 × 10 <sup>-6</sup>
BUM	B.2	B.2	B.1	B.2	no	B.3

Table 1: Model main characteristics.  $d$  is water depth and  $k_2$  is the <sup>137</sup>Cs desorption coefficient. <sup>a</sup> See for instance Cushman-Roisin and Beckers (2011). A selected reference is given for each model. The BUM row indicates the appendix where some details of the uptake model are given: B.1 is an equilibrium model, B.2 is a dynamic model and B.3 is the allometric method.

149 piscivorous fish) in the surface layer (to 20 m depth). The other two considered water  
150 layers are 20-460 m and 460 m to the seabed.

151 Measurements were compiled from the following references: Honda et al. (2012),  
152 Charette et al. (2013), Kaeriyama et al. (2013) for water; the “Database for Radioac-  
153 tive Substance Monitoring Data”<sup>4</sup> for sediments; Honda et al. (2012), Kitamura et al.  
154 (2013) for zooplankton; Wada et al. (2016); Men et al. (2017); Johansen et al. (2014)  
155 for fish (piscivorous and non-piscivorous). Only data for pelagic fish were used. Sam-  
156 pled pelagic non-piscivorous fish are *Engraulis japonicus*, *Etrumeus teres*, *Clupea pallasii*  
157 and *Hyporhamphus sajori*. Sampled pelagic piscivorous fish are *Hexagrammos seabastes*,  
158 *Todarodes pacificus*, *Snake mackerel*, *Oncorhynchus keta*, *Ammodytes japonicus*, *Seriola*  
159 *quinqueradiata*, *Seriola quinqueradiata*, *Trachurus japonicus* and *Scomber japonicus*. Wa-  
160 ter samples collected in the direct release area have been filtered out since the models are  
161 giving average value of radionuclide concentrations over boxes, as explained below.

162 Locations where samples were collected during the simulation period are indicated as  
163 dots in Fig. 3. The Pacific Ocean was divided into a number of boxes, presented in Fig. 4,  
164 according to general circulation and the location of the release point. Model results were  
165 averaged for each box and then these averaged values were compared with measurements.

166 Boxes in the release area may be too large for a detailed study of radionuclide be-  
167 haviour in such region close to FDNPP. However, it should be taken into account that  
168 the dispersion of FDNPP <sup>137</sup>Cs releases was studied at a smaller spatio-temporal scale in  
169 a previous paper of the group (Periáñez et al., 2015a); and model predictions and mea-  
170 surements were compared in the area close to FDNPP (less than some 100 km away).  
171 The present work is complementing such previous paper, going to larger spatial and tem-  
172 poral scales. Thus, large boxes are used. In addition, it should be considered that a  
173 model/data comparison for specific points in such a large domain is not feasible with  
174 Lagrangian models which release individual particles, and it is better to use averages over

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<sup>4</sup><http://emdb.jaea.go.jp/emdb/en/>



175 given areas, which are defined in view of the physical oceanography of the region (Periáñez  
176 et al., 2015a; 2015b; 2016a). However, it should be noted that measurements were not  
177 distributed homogeneously in the relatively large considered boxes.

## 178 **3 Results**

179 As explained before, two year long simulations were carried out; from March 2011 to  
180 March 2014. Monthly mean values of  $^{137}\text{Cs}$  concentrations in each box in Fig 4 were  
181 provided by the models for the three water layers, seabed sediments and the four biological  
182 compartments (surface layer only).

183 Model results and  $^{137}\text{Cs}$  measurements are presented in Fig. 5 to Fig. 12. Results are  
184 presented only for such boxes where measurements are available. Results for the abiotic  
185 and biotic components of the models are discussed separately in the following subsections.

### 186 **3.1 Water and sediments**

187 Results for surface water may be seen in Fig. 5 and Fig. 6, for boxes which are far  
188 from Japan and boxes located closer, around FDNPP, respectively. In boxes 1, 3, 5 and  
189 20 (Fig. 5) there is a slight increase in  $^{137}\text{Cs}$  concentrations with respect to background  
190 immediately after the accident, which must be attributed to atmospheric deposition. In  
191 general, models produce this initial increase, which is about one order of magnitude above  
192 background. In other boxes (like 15 and 16), both models and measurements indicate pre-  
193 FDNPP accident background. Thus, releases did not affect these areas in the considered  
194 temporal frame.

195 In contrast, high concentrations are found closer to FDNPP (Fig. 6). For some of the  
196 boxes (6, 7, 12) models and measurements show a trend towards achieving background  
197 concentrations after approximately one year. The initial concentration increase above  
198 background is about two orders of magnitude. Other regions south from Japan (boxes 13

199 and 14) do not seem to be affected by FDNPP releases, although some models predict  
200 a slight concentration increase above background. Direct releases from FDNPP into the  
201 ocean occur in box 11; consequently very high  $^{137}\text{Cs}$  concentrations were measured here,  
202 especially soon after the accident. Although samples collected just in the release area are  
203 not considered, as commented above, obviously there must be a significant underestima-  
204 tion of  $^{137}\text{Cs}$  concentrations in this box in the period of acute discharges. This can be  
205 clearly seen in Fig. 7, where the geometric means of measured concentrations in boxes  
206 10 and 11 for each month are represented together with box averaged model predictions.  
207 Mean values of measured concentrations in both boxes decrease in up to three orders of  
208 magnitude. While all models underestimate measurements in the early period after the  
209 accident, the mean values of measurements are within the range of model predictions after  
210 such initial phase.

211 Generally speaking, models agree in predicting areas in the Pacific Ocean which were  
212 affected by FDNPP releases (direct and/or atmospheric deposition) and regions which  
213 were not. In addition, there is a relatively good agreement between the temporal trends  
214 of  $^{137}\text{Cs}$  concentrations predicted by the different models. In most cases, concentrations  
215 are within the same order of magnitude.

216 Results for intermediate and deep waters are presented in Fig. 8. In these cases  
217 measurements were not carried out soon after the accident. Models indicate an increase  
218 in  $^{137}\text{Cs}$  concentrations (about a factor  $10^2$ ) immediately after FDNPP accident and a  
219 decrease to background concentrations for intermediate waters. Both models and mea-  
220 surements show concentrations in the deep later of the same order of magnitude as back-  
221 ground. Thus, deep water in these regions (boxes 10 and 11) was not affected by the  
222 accident in the considered temporal frame. Again, there is a relatively good agreement  
223 between outputs of the different models.

224 The case of sediments may be seen in Fig. 9.  $^{137}\text{Cs}$  concentration in the bed sediment  
225 increase significantly over background in boxes 10 and 11, which are the closer to FDNPP

226 than box 9. Models predict such increase, although with larger discrepancy between  
227 models than in the case of the dissolved phase. In contrast, background levels are apparent  
228 in box 9, north from the previous ones. It is interesting to note that a decreasing trend of  
229  $^{137}\text{Cs}$  concentrations in sediments close to FDNPP (less than some 100 km away) has been  
230 observed (Kusakabe et al., 2013) after the period of acute releases. Sediments at larger  
231 distances from FDNPP do not show such decreasing trend (dots in Fig. 9, box 10 and  
232 11). Thus, it seems that these sediments are buffering radionuclides. Most models also  
233 predict quite constant  $^{137}\text{Cs}$  concentrations in these sediments and the range of model  
234 predictions is within the range of measured concentrations. The geometric means of  
235 measured concentrations in boxes 10 and 11 for each month are represented together with  
236 box averaged model predictions in Fig. 7. Mean values of measured concentrations in  
237 sediments in both boxes effectively do not show any clear decreasing trend; and these  
238 mean values of measurements are within the range of model predictions, as mentioned  
239 above.

### 240 **3.2 Biotic components**

241 Results for zooplankton are presented in Fig. 10 for boxes where measurements are avail-  
242 able. Model results are in general within the same order of magnitude. However, most  
243 models significantly underestimate measured concentrations. Zooplankton takes  $^{137}\text{Cs}$   
244 from phytoplankton (which in dynamic models is in equilibrium with water [see appendix  
245 B.2]) and directly from water as well. Thus, there are not significant differences in the  
246 temporal trends produced for zooplankton by equilibrium and dynamic models. The allo-  
247 metric approach (LORAS model) leads to larger concentrations in boxes 7 and 12, which  
248 are in better agreement with observations than the other models. The better agreement of  
249 LORAS model with measurements for zooplankton (and further for fish) can be explained  
250 by the higher concentration in water predicted by this model.

251 Non zero concentrations in biota calculated by USEV model before FDNPP accident  
252 are due to the fact that background concentrations are assumed in water since  $t = 0$ ,  
253 that in this model is January 1st. Thus, biotic components of the model absorb such  
254 background  $^{137}\text{Cs}$ .

255 Results for non-piscivorous and piscivorous fish are respectively shown in Figs. 11  
256 and 12. Now differences between equilibrium and dynamic models become clearer.  $^{137}\text{Cs}$   
257 concentrations predicted by an equilibrium model reflect concentrations in water. Thus,  
258 peak values are reached simultaneously in water and fish. There is a delay in the time  
259 of peak fish concentrations with respect to peak concentrations in water in the case of  
260 dynamic models. This different behaviour of equilibrium and dynamic models was already  
261 pointed out in the BUM intercomparison carried out by Vives i Battle et al. (2016) in the  
262 frame of IAEA MODARIA program. However, in such exercise BUMs were not included  
263 within full marine dispersion models. Instead, BUM model responses were tested in a  
264 single point where  $^{137}\text{Cs}$  concentration in water was prescribed (what could be denoted  
265 as a zero-dimensional model).

266 Both models including a dynamic BUM (THREETOX and USEV) produce similar  
267 outputs in spite of being models with different structure (Eulerian and Lagrangian re-  
268 spectively). Both models use the same generic parameters described in Maderich et al.  
269 (2014a). Thus, these results are supporting the findings in Maderich et al. (2018): this  
270 BUM model is robust enough to be used with generic parameters in areas where limited  
271 information is available. Concentration levels produced by the equilibrium model in fish  
272 are, in general, in better agreement with measurements than calculations by dynamic  
273 models and allometry. This is due to the higher concentrations that ESTE model (which  
274 uses the equilibrium approach) produces in water (Fig. 6) and should not be attributed to  
275 the BUM itself. Nevertheless, the instantaneous equilibrium with activity concentration  
276 in seawater does not seem to be realistic (Vives i Battle et al., 2016).

277 The allometric model output generally is more similar to the dynamic models in the

278 case of non-piscivorous fish and to the equilibrium model in the case of piscivorous fish.  
279 However there are not enough experimental data to assess which model produces a more  
280 realistic behaviour.

## 281 **4 Discussion**

282 In a previous work (Periáñez et al., 2015a) it was found that the main source of dis-  
283 crepancy between dispersion models is water circulation. Discrepancy between models  
284 is significantly reduced if the same water circulation fields are used by all dispersion  
285 models. However, other sources of uncertainty in model/model comparisons exist. This  
286 uncertainty is due to model parameters and model numerics.

287 If the model is applied to a perfectly conservative radionuclide (remaining dissolved,  
288 without any interactions with sediments) the only involved parameters are the horizontal  
289 and vertical diffusion coefficients. Turbulence is an open problem in physics and has to  
290 be parameterized. Thus, different schemes and approaches are used to evaluate diffusion  
291 coefficients. These different approaches may lead to different model results. The situation  
292 is even more complex for non-conservative radionuclides. A number of parameters is  
293 required in this case, like kinetic rates, particle sizes, density and thickness of the sediment,  
294 etc. These parameters are site-specific and information about them is generally scarce.  
295 Thus, only realistic or tentative values have often to be used. Dynamic models need more  
296 parameters in comparison with more simple approaches based on sediment distribution  
297 coefficients and concentration ratios for biota, but they are less site-dependent.

298 The second source of uncertainty is due to model numerics. A numerical solution  
299 always requires discretization in time. A spatial discretization is required in Eulerian  
300 models. However, even in Lagrangian models a spatial discretization is required when  
301 concentrations are derived from the number of particles per water volume unit, resulting  
302 in averaging of quantities. A discretization always implies averaging magnitudes; and

303 averaging leads to errors. Moreover, a numerical solution is only an approximation to the  
304 exact solution since several errors appear (rounding errors, truncation errors etc). The  
305 radionuclide release area size has to be considered as well. In Eulerian models, radionu-  
306 clides are homogeneously distributed into the release cell where the accident occurs; this  
307 would be the initial patch minimum size. Thus, the initial patch size depends on the  
308 model spatial resolution. This initial patch size defines the initial peak concentration. In  
309 contrast, a real point source can be used in a Lagrangian model. Obviously, this will lead  
310 to differences between Eulerian and Lagrangian models.

311 It is essential to have accurate oceanographic data to compare model results with  
312 measurements. However, it is hard to obtain accurate predictions of water currents in  
313 energetic regions characterized by strong current variability, like Fukushima waters and  
314 the North Western Pacific region, which are characterized by the very strong and fluctuat-  
315 ing Kuroshio current and its extension (Masumoto et al., 2012). Different hydrodynamic  
316 models will lead to slightly different current fields. Given the intensity and variability of  
317 currents in these energetic areas, as well as the presence of unsteady eddies, small differ-  
318 ences in the hydrodynamics may produce differences in dispersion patterns which tend to  
319 be amplified with time.

320 Another issue is how to choose the most appropriate level of model complexity for a  
321 given problem. Of course, this is related to the expected end-point of a simulation. This  
322 will define the spatio-temporal resolving power of the model (Monte et al., 2006), which is  
323 a measure of the level of detail of its predictions. The time resolving power is the ability  
324 of a model to predict differences in the system behaviour over a given interval of time.  
325 Similarly, the spatial resolving power is the ability of a model to predict differences in the  
326 system behaviour over a given spatial grid. Of course, model complexity increases with  
327 the resolving power.

328 Models with different complexities have been used in the present paper to calculate  
329 radionuclide concentrations in biota (from equilibrium to dynamic models). As stated in

330 Monte et al. (2006), the general principle that the simplest model is ever better than  
331 the complex one, if they supply similar results for some given particular applications,  
332 should be avoided. A simple model may not be sufficiently developed for application  
333 to the innumerable possible contamination scenarios, marine systems and environmental  
334 circumstances that other more complex and general models are meant to simulate. For  
335 instance, as has been commented above, peak  $^{137}\text{Cs}$  concentrations are reached simulta-  
336 neously in water and fish if an equilibrium model is used. On the other hand, there is a  
337 delay in the time of peak fish concentrations with respect to peak concentrations in water  
338 in the case of dynamic models; which is more realistic. Thus, the simple (equilibrium)  
339 model is not suitable for the study of the initial phase of an accident. A more complex  
340 model, able to deal with non-equilibrium situations, should be applied for this purpose.

## 341 5 Conclusions

342 A number of physical dispersion models of different natures (Eulerian and Lagrangian)  
343 were applied to simulate FDNPP  $^{137}\text{Cs}$  releases in the Pacific Ocean over two years and at  
344 oceanic scale. Realistic source terms for direct releases and atmospheric deposition were  
345 used. Most models included a biological uptake model consisting of four species: phy-  
346 toplankton, zooplankton, non-piscivorous and piscivorous fish. There types of biological  
347 uptake models were tested: equilibrium model, dynamic model and an allometry method.  
348 Model results were compared with measurements in water (three layers), sediment and  
349 biota.

350 In general, there is a good agreement between models and between models and mea-  
351 surements. The method used to compare models is helping in this regards. It was found  
352 (Periáñez et al., 2015a) that comparisons of Lagrangian model outputs in specific points  
353 is difficult since a number of discrete particles are released in these models. Thus, it is  
354 more convenient to divide the oceanic space into a number of boxes and to obtain average

355 concentrations over such boxes; as it was done for the Baltic Sea model intercomparison  
356 in Perri  ez et al. (2015b). However, it should be taken in account that in the vicinity of  
357 FDNPP measurements were not distributed homogeneously in space. This can result in  
358 overestimation of experimental box-averaged values.

359 Models agree in predicting areas in the Pacific Ocean which were affected by FDNPP  
360 releases (direct and/or atmospheric deposition) and regions which were not. In addition,  
361 predicted concentrations are within the same order of magnitude in most cases.

362 With respect to calculated  $^{137}\text{Cs}$  temporal trends in biota, dynamic models tend to  
363 underestimate concentrations. Allometry and the equilibrium approach results are, in  
364 general, in better agreement with observations. This is explained by the higher  $^{137}\text{Cs}$   
365 concentrations in water produced by ESTE and LORAS models. Temporal evolutions of  
366  $^{137}\text{Cs}$  concentrations calculated through the different approaches are different, although  
367 there is not enough experimental data to assess which approach leads to better results.  
368 However, it is clear that dynamic models provide the known pattern of delayed rise of  
369 activity concentration in biota.

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## 6 References

Bezhenar, R., Jung, K.T., Maderich, V., Willemsen, S., de With, G., Qiao, F., 2016. Transfer of radiocesium from contaminated bottom sediments to marine organisms through benthic food chain in post-Fukushima and post-Chernobyl periods. *Biogeosciences* 13, 3021-3034.

Brovchenko, I., Jung, K.T., Maderich, V., Kim, K.O., Kovalets, S., 2018. Method of moments for random walk algorithms. In preparation.

Brown, J., Borrentzen, P., Dowdall, M., Sazykina, T., Kryshev, I., 2004. The Derivation of Transfer Parameters in the Assessment of Radiological Impacts on Arctic marine Biota. *ARCTIC* 57(3), 279-289.

Carvalho, F. P., 2018. Radionuclide concentration processes in marine organisms: A comprehensive review. *Journal of Environmental Radioactivity* 186, 124-130.

Charette, A.M. Breier, C.F., Henderson, P.B., Pike, S.M., Rypina, I.I., Jayne, S.R., Buesseler, K.O., 2013. Radium-based estimates of cesium isotope transport and total direct ocean discharges from the Fukushima Nuclear Power Plant accident, *Biogeosciences* 10, 2159-2167.

Cushman-Roisin, B., Beckers, J.M., 2011. *Introduction to Geophysical Fluid Dynamics*. Elsevier.

Heling, R., Koziy, L., Bulgakov, V., 2002. On the dynamical uptake model developed for the uptake of radionuclides in marine organisms for the POSEIDON-R model system. *Radioprotection* 37(C1), 833-838.

Higley, K.A., Domotor, S.L., Antonio, E.J., Kocher, D.C., 2003. Derivation of a screening methodology for evaluating radiation dose to aquatic and terrestrial biota. *Journal of Environmental Radioactivity*, 66, 41-59.

401 Honda M., Aono T., Aoyama M., Hamajima Y., Kawakami H., Kitamura M., Ma-  
402 sumoto Y., Miyazawa Y., Takigawa M., Saino T., 2012. Dispersion of artificial  
403 caesium-134 and -137 in the western North Pacific one month after the Fukushima  
404 accident. *Geochemical Journal* 46, 1-9.

405 IAEA, 2004. Sediment distribution coefficients and concentration factors for biota  
406 in the marine environment. Technical Reports Series 422, Vienna.

407 Johansen, M.P., Ruedig, E., Tagami, K., Uchida, S., Higley, K., Beresford, N., 2014.  
408 Radiological dose rates to marine fish from the Fukushima Daiichi accident: the first  
409 three years across the North Pacific, *Environmental Science and Technology* 49(3),  
410 1277-1285.

411 Kaeriyama, H., Ambe, D., Shimizu, Y., Fujimoto, K., Ono, T., Yonezaki, S., Kato,  
412 Y., Matsunaga, H., Minami, H., Nakatsuka, S., Watanabe, T., 2013. Direct ob-  
413 servation of  $^{134}\text{Cs}$  and  $^{137}\text{Cs}$  in surface seawater in the western and central North  
414 Pacific after the Fukushima Dai-ichi nuclear power plant accident, *Biogeosciences*  
415 10, 4287-4295.

416 Kawamura, H., Furuno, A., Kobayashi, T., In, T., Nakayama, T., Ishikawa, Y.,  
417 Miyazawa, Y., Usui, N., 2017. Oceanic dispersion of Fukushima-derived Cs-137  
418 simulated by multiple oceanic general circulation models. *Journal of Environmental*  
419 *Radioactivity* 180, 36-58.

420 Kitamura, M., Kumamoto, Y., Kawakami, H., Cruz, E.C., Fujikura, K., 2013. Hor-  
421 izontal distribution of Fukushima-derived radiocesium in zooplankton in the north-  
422 western Pacific Ocean. *Biogeosciences* 10, 5729-5738.

423 Kobayashi, T., Otosaka, S., Togawa, O., Hayashi, K., 2007. Development of a  
424 nonconservative radionuclides dispersion model in the ocean and its application

425 to surface cesium-137 dispersion in the Irish Sea. *Journal of Nuclear Science and*  
426 *Technology* 44(2), 238-247.

427 Kobayashi, T., Nagai, H., Chino, M., Kawamura, H., 2013. Source term estimation  
428 of atmospheric release due to the Fukushima Dai-ichi Nuclear Power Plant accident  
429 by atmospheric and oceanic dispersion simulations. *Journal of Nuclear Science and*  
430 *Technology* 50, 255-264.

431 Kusakabe, M., Oikawa, S., Takata, H., Misonoo, J., 2013. Spatiotemporal dis-  
432 tributions of Fukushima-derived radionuclides in nearby marine surface sediments.  
433 *Biogeosciences* 10, 5019-5030.

434 Lepicard S., Heling R., Maderich V. 2004. POSEIDON/RODOS model for radio-  
435 logical assessment of marine environment after accidental releases: application to  
436 coastal areas of the Baltic, Black and North seas. *Journal of Environmental Ra-*  
437 *dioactivity* 72, 1-2, 153-161

438 Lynch D.R., Greenberg D.A., Bilgili A., Mcgillicuddy, D.J., Manning, J.P., Aretx-  
439 abaleta A.L., 2015. *Particles in the Coastal Ocean. Theory and Applications.*  
440 Cambridge University Press, NY.

441 Maderich V., Heling R., Bezhenar R., Brovchenko I., Jenner H., Koshebutskyy V.,  
442 Kuschak A., Terletska K.. 2008. Development and application of 3D numerical  
443 model THREETOX to the prediction of cooling water transport and mixing in the  
444 inland and coastal waters. *Hydrological Processes* 22, 1000-1013.

445 Maderich, V., Bezhenar, R., Heling, R., de With, G., Jung, K.T., Myoung, J.G.,  
446 Cho, Y.K., Qiao, F., Robertson, L., 2014a. Regional long-term model of radioac-  
447 tivity dispersion and fate in the northwestern Pacific and adjacent seas: application  
448 to the Fukushima Dai-ichi accident. *Journal of Environmental Radioactivity* 131,  
449 4-18.

450 Maderich V., Jung, K.T., Bezhenar R., de With, G., Qiao F., Casacuberta, N.,  
451 Masqué, P., Kim, Y.H., 2014b. Dispersion and fate of  $^{90}\text{Sr}$  in the Northwestern Pa-  
452 cific and adjacent seas: Global fallout and the Fukushima Dai-ichi accident. *Science*  
453 *of the total Environment* 494-495, 261-271.

454 Maderich V., Brovchenko I., Dvorzhak A., Ievdin I., Koshebutsky, V., Periañez, R.,  
455 2016. Integration of 3D model THREETOX in JRODOS, implementation studies  
456 and modelling of Fukushima scenarios. *Radioprotection* 51(HS2), S133-S135.

457 Maderich, V., Jung, K.T., Brovchenko, I., Kim, K.O., 2017. Migration of radioac-  
458 tivity in multi-fraction sediments. *Environmental Fluid Mechanics* 17, 1207-1231.

459 Maderich, V., Bezhenar, R., Tateda, Y., Aoyama, M., Tsumune, D., 2018. Sim-  
460 ilarities and differences of  $^{137}\text{Cs}$  distributions in the marine environments of the  
461 Baltic and Black seas and off the Fukushima Dai-ichi nuclear power plant in model  
462 assessments. *Marine Pollution Bulletin* 135, 895-906.

463 Masumoto, Y., Miyazawa, Y., Tsumune, D., Kobayashi, T., Estournel, C., Marsaleix,  
464 P., Lanerolle, L., Mehra, A., and Garraffo, Z.D., 2012. Oceanic dispersion simulation  
465 of Cesium-137 from Fukushima Dai-ichi nuclear power plant, *Elements*, 8, 207-212.

466 Men, W., Deng, F., He, J., Yu, W., Wang, F., Li, Y., Lin, F., Lin, J., Lin, L., Zhang,  
467 Y., Yu, X., 2017. Radioactive impacts on nekton species in the Northwest Pacific  
468 and humans more than one year after the Fukushima nuclear accident, *Ecotoxicology*  
469 *and Environmental Safety* 144, 601-610.

470 Min, B.I., Periañez, R., In-Gyu Kim, Kyung-Suk Suh, 2013. Marine dispersion  
471 assessment of  $^{137}\text{Cs}$  released from the Fukushima nuclear accident. *Marine Pollution*  
472 *Bulletin* 72, 22-33.

473 Monte, L., Hakason, L., Periañez, R., Laptev, G., Zheleznyak, M., Maderich, V.,

474 Koshebutsky, V., 2006. Experiences from a case study of multi-model application to  
475 assess the behaviour of pollutants in the Dnieper-Bug estuary. *Ecological Modelling*  
476 195, 247-263.

477 Nyffeler, U.P., Li, Y.H., Santschi, P.H., 1984. A kinetic approach to describe  
478 trace element distribution between particles and solution in natural aquatic sys-  
479 tems. *Geochimica Cosmochimica Acta* 48, 1513-1522.

480 Periañez, R., Elliott, A. J., 2002. A particle-tracking method for simulating the  
481 dispersion of non-conservative radionuclides in coastal waters. *Journal of Environ-*  
482 *mental Radioactivity* 58(1), 13-33.

483 Periañez, R., 2005. Modelling the dispersion of radionuclides in the marine environ-  
484 ment: an introduction. Springer, Berlin.

485 Periañez, R., Bezhenar, R., Byung-Il, Min, Duffa, C., Jung, K., Kobayashi, T.,  
486 Kyung-Suk, Suh, Lamego, F., Maderich, V., Nies, H., Osvath, I., Psaltaki, M., 2015a.  
487 A new comparison of marine dispersion model performances for Fukushima releases  
488 in the frame of IAEA MODARIA program. *Journal of Environmental Radioactivity*  
489 150, 247-269.

490 Periañez, R., Bezhenar, R., Iosjpe, M., Maderich, V., Nies, H., Osvath, I., Outola, I.,  
491 de With, G., 2015b. A comparison of marine radionuclide dispersion models for the  
492 Baltic Sea in the frame of IAEA MODARIA program. *Journal of Environmental*  
493 *Radioactivity* 139, 66-77.

494 Periañez, R., R. Bezhenar, I. Brovchenko, C. Duffa, K.T. Jung, T. Kobayashi, F.  
495 Lamego, V. Maderich, B.I. Min, H. Nies, I. Osvath, I. Outola, M. Psaltaki, K.S.  
496 Suh, G. de With, 2016a. Modelling of marine radionuclide dispersion in IAEA  
497 MODARIA program: lessons learnt from the Baltic Sea and Fukushima scenarios.  
498 *Science of the Total Environment* 569/570, 594-602.

499 Periañez, R., Kyung-Suk Suh, Byung-Il Min, 2016b. The behaviour of  $^{137}\text{Cs}$  in the  
500 North Atlantic Ocean assessed from numerical modelling: releases from nuclear fuel  
501 reprocessing factories, redissolution from contaminated sediments and leakage from  
502 dumped nuclear wastes. *Marine Pollution Bulletin* 113, 343-361.

503 Proehl, J.A., Lynch, D.R., McGuillicuddy, D.J., Ledwell, J.R., 2005. Modeling  
504 turbulent dispersion on the North Flank of Georges Bank using Lagrangian particle  
505 methods. *Continental Shelf Research* 25, 875-900.

506 Protter, P.E., 2004. *Stochastic Integration and Differential Equations* (2nd ed.).  
507 Springer, ISBN 3-540-00313-4.

508 Schmidt-Nielsen, K., 1977. Problems of scaling: locomotion and physiological cor-  
509 relates. In: *Scale Effects in Animal Locomotion*, T.J. Pedley (editor). Academic  
510 Press, New York. pp. 1-21.

511 SCJ, 2014. A review of the model comparison of transportation and deposition of  
512 radioactive materials released to the environment as a result of the Tokyo Electric  
513 Power Compny's Fukushima Daiichi Nuclear Power Plant accident. Report of the  
514 Sectional Committee on Nuclear Accident Committee on Comprehensive Synthetic  
515 Engineering, Science Council of Japan.

516 Suh, K.S., Jeong, H.J., Kim, E.H., Hwang, W.T., Han, M.H., 2006. Verification  
517 of the Lagrangian particle model using the ETEX experiment. *Annals of Nuclear*  
518 *Energy* 33, 1159-1163.

519 Suh, K.S., Han, M.H., Jung, S.H., Lee, C.W., 2009. Numerical simulation for a long-  
520 range dispersion of a pollutant using Chernobyl data. *Mathematical and Computer*  
521 *Modeling* 49, 337-343.

522 Takemura, T., Nakamura, H., Takigawa, M., Kondo, H., Satomura. T., Miyasaka,

523 T., Nakajima, T., 2011. A numerical simulation of global transport of atmospheric  
524 particles emitted from the Fukushima Daiichi Nuclear Power Plant. SOLA 7, 101-  
525 104.

526 TEPCO, 2011. <http://www.tepco.co.jp/en/nu/fukushima-np/f1/index2-e.html>.

527 Terada, H., Katata, G., Chino, M., Nagai, H., 2012. Atmospheric discharge and  
528 dispersion of radionuclides during the Fukushima Dai-ichi Nuclear Power Plant ac-  
529 cident. Part II: verification of the source term and analysis of regional-scale atmo-  
530 spheric dispersion. *Journal of Environmental Radioactivity* 112, 141-154.

531 Tsujino, H., Motoi, T., Ishikawa, I., Hirabata, M., Nakano, H., Yamanaka, G.,  
532 Yasuda, T., Ishizaki, H., 2010. Reference Manual for the Meteorological Research  
533 Institute Community Ocean Model (MRI.COM) Version 3. Tsukuba (Japan): Me-  
534 teorological Research Institute. Technical Reports of the MRI, 59.

535 Usui, N., Wakamatsu, T., Tanaka, Y., Hirose, N., Toyoda, T., Nishikawa, S., Fujii,  
536 Y., Takatsuki, Y., Igarashi, H., Ishikawa, Y., Kuragano, T., Kamachi, M., 2017.  
537 Four-dimensional Variational Ocean Reanalysis: A 30-year high-resolution dataset  
538 in the western North Pacific (FORA-WNP30). *Journal of Oceanography* 73(2),  
539 205-233.

540 Vives i Batlle, J., Wilson, R.C., Watts, S.J., Jones, S.R., McDonald, P., Vives-  
541 Lynch, S., 2008. Dynamic model for the assessment of radiological exposure to  
542 marine biota, *Journal of Environmental Radioactivity* 99, 1711-1730.

543 Vives i Batlle, J., Beresford, N., Beaugelin-Seiller, K., Bezhenar, R., Brown, J., Jing-  
544 Jy Cheng, Cujic, M., Dragovic, S.S., Duffa, C., Fievet, B., Hosseini, A., Jung, K.T.,  
545 Kamboj, S., Dong-Kwon Keum, Kobayashi, T., Kryshev, A., Le Poire, D., Maderich,  
546 V., Byung Il Min, Periañez, R., Sazykina, T., Suh, K.S., Yu, C., Wang, C., Heling,

547 R., 2016. Inter-comparison of dynamic models for radionuclide transfer to marine  
548 biota in a Fukushima accident scenario. *Journal of Environmental Radioactivity*  
549 153, 31-50.

550 Wada, T., Fujita, T., Nemoto, Y., Shimamura, S., Mizuno, T., Sohtome, T., Kami-  
551 yama, K., Narita, K., Watanabe, M., Hatta, N., Ogata, Y., Morita, T., Igarashi, S.,  
552 2016. Effects of the nuclear disaster on marine products in Fukushima: An update  
553 after five years. *Journal of Environmental Radioactivity* 164, 312-324.

554 West G.B., Brown, J.H., Enquist, B.J., 1997. A general model for the origin of  
555 allometric scaling laws in biology. *Science* 276, 122-126.

## 556 **A Physical dispersion models**

557 Physical dispersion models (both Eulerian and Lagrangian) are based on the same general  
558 principles and equations; and then particularized as presented in Table 1. Thus, those  
559 common general descriptions are given below.

### 560 **A.1 Eulerian models**

561 In Eulerian models the differential equations giving temporal and spatial evolution of the  
562 radionuclide concentrations in different states (e.g. dissolved in water column and pore  
563 water in sediments, fixed on the suspended and bottom sediment etc) are solved. The  
564 general compact form of these equations for concentration of radioactivity  $C_\alpha$  in state  
565  $\alpha$  per unit of volume ( $\text{Bq m}^{-3}$ ) or per unit of mass ( $\text{Bq kg}^{-1}$ ) are written in Cartesian  
566 coordinates as:

$$\frac{\partial C_\alpha}{\partial t} + \frac{\partial(u_\alpha C_\alpha)}{\partial x} + \frac{\partial(v_\alpha C_\alpha)}{\partial y} + \frac{\partial(w_\alpha C_\alpha)}{\partial z} = \frac{\partial}{\partial x} \left( K_h \frac{\partial C_\alpha}{\partial x} \right) + \frac{\partial}{\partial y} \left( K_h \frac{\partial C_\alpha}{\partial y} \right) +$$



$$+\frac{\partial}{\partial z} \left( K_v \frac{\partial C_\alpha}{\partial z} \right) + \sum_{\beta=1}^n k_{\beta\alpha} C_\beta + S_\alpha - \lambda C_\alpha \quad (1)$$

567 where  $(x, y, z)$  are Cartesian coordinates,  $u_\alpha$ ,  $v_\alpha$  and  $w_\alpha$  are components of flow field for the  
 568 radionuclide in the state  $\alpha$ . In general, velocity can differ for different states (e.g. due the  
 569 presence of settling velocity for suspended sediment or to be zero in the bottom deposit).  
 570  $K_h$  and  $K_v$  are turbulent or molecular diffusivities in the horizontal and vertical directions  
 571 respectively, and/or biodiffusivity in the bottom deposit, which are variable in time and  
 572 space. The term  $\sum_{\beta=1}^n k_{\beta\alpha} C_\beta$  describes first order reactions between the radionuclides in  
 573 different states,  $k_{\beta\alpha}$  are kinetic transfer coefficients and  $k_{\alpha\alpha} = -\sum_{\beta=1}^n k_{\alpha\beta}$  for  $\alpha \neq \beta$ ;  $S_\alpha$   
 574 is the radionuclide source term and  $\lambda$  is the radionuclide decay constant. Equations for  
 575 the water column and bottom sediment layer are linked by fluxes of activity.

## 576 A.2 Lagrangian models

577 In Lagrangian models the released activity is represented by a number of particles, each  
 578 one equivalent to a given amount of activity (Bq). The path followed by each particle is  
 579 calculated and radionuclide concentrations are obtained from the number of particles per  
 580 volume or mass unit. The equations describing variations of particle (in state  $\alpha$ ) position  
 581 over each time increment  $dt$  are given by the Itô (Protter, 2004) stochastic differential  
 582 equations:

$$dx = u_\alpha dt + \frac{\partial K_h}{\partial x} dt + \sqrt{2K_h} dW_x, \quad (2)$$

$$dy = v_\alpha dt + \frac{\partial K_h}{\partial y} dt + \sqrt{2K_h} dW_y, \quad (3)$$

$$dz = w_\alpha dt + \frac{\partial K_v}{\partial z} dt + \sqrt{2K_v} dW_z, \quad (4)$$

583 where  $u_\alpha$ ,  $v_\alpha$  and  $w_\alpha$  are velocity components on coordinate axis  $(x, y, z)$  for state  $\alpha$ ;  
 584  $W_x, W_y, W_z$  are independent components of the stochastic motion (the Wiener process).

585 They have zero mean and variance  $dt$  ( $\overline{dW_x^2} = \overline{dW_y^2} = \overline{dW_z^2} = dt$ ). For the finite time  
586 step  $\Delta t$  Wiener increments can be simulated as  $\Delta W_x = \sqrt{\Delta t}R_x$ ,  $\Delta W_y = \sqrt{\Delta t}R_y$ ,  $\Delta W_z =$   
587  $\sqrt{\Delta t}R_z$ , where  $(R_x, R_y, R_z)$  are normally distributed random variables having zero mean  
588 and standard deviation one. Derivatives of the diffusion coefficients above prevent the  
589 artificial accumulation of particles in regions of low diffusivity (Proehl et al., 2005; Lynch  
590 et al., 2015).

591 An stochastic method is used to solve the exchanges of radionuclides between the liquid  
592 and solid phases in a dynamic way. These processes are formulated using kinetic transfer  
593 coefficients, considering that exchanges of radionuclides between phases are governed by  
594 a first-order reversible reaction. Detail may be seen in Perri  nez and Elliott (2002).

## 595 **B Biological uptake models**

596 Three types of biological uptake models have been applied. An equilibrium model based  
597 on concentration ratios, a dynamic model and an allometry method. They are briefly  
598 described below.

### 599 **B.1 Equilibrium model**

600 The equilibrium approximation is based on a concentration factor,  $CR$ , between water and  
601 biota. This  $CR$ , in analogy with the water/sediment  $k_d$ , is defined as the ratio between  
602 radionuclide concentration in a given specie of biota and concentration in water:

$$CR = \frac{C_{bio}}{C_w}. \quad (5)$$

603 Thus, concentration in biota,  $C_{bio}$ , can be calculated from the  $CR$  and the calculated  
604 concentration in water, assuming equilibrium (Carvalho, 2018). Concentration factors are  
605 determined for fishes, mollusks, crustaceans, plankton and algae, and for a large number of

606 elements (IAEA, 2004). These generic values have been used. In the equilibrium approach,  
 607 used by ESTE model, non-piscivorous and piscivorous fish are not distinguished. Thus,  
 608 results for both types of fish are supposed to be the same.

## 609 **B.2 Dynamic model**

610 The dynamic model consists of four species (Heling et al., 2002; Maderich et al., 2014a;  
 611 2014b): phytoplankton, zooplankton, non-piscivorous and piscivorous fish (Fig. 13, from  
 612 Maderich et al., 2014a). The basic equation connecting concentration of activity in preda-  
 613 tor  $C_{pred}$  (Bq kg<sup>-1</sup> wet weight) with activity concentration in food  $C_f$  (Bq kg<sup>-1</sup> wet weight)  
 614 is:

$$\frac{\partial C_{pred}}{\partial t} = aK_1C_f + bK_wC_w - K_{0.5}C_{pred}, \quad (6)$$

615 where  $K_1$  ( $s^{-1}$ ) is food uptake rate,  $a$  is the transfer coefficient through food,  $K_w$  is water  
 616 uptake rate ( $s^{-1}$ ),  $b$  is the transfer coefficient from water and  $C_w$  is activity concentration  
 617 in water (Bq m<sup>-3</sup>).  $K_{0.5}$  is the radionuclide elimination rate from the body of fish given  
 618 by  $K_{0.5} = \ln 2T_{0.5}^{-1}$ , where  $T_{0.5}$  is the biological half-life of the radionuclide (s). Thus,  
 619 all organisms take radionuclides from water, phytoplankton is the food for zooplankton,  
 620 zooplankton is the food for non-piscivorous fish and this is the food for piscivorous fish (as  
 621 summarized in Fig. 13). Phytoplankton exchanges radionuclides only with the water via  
 622 adsorption and desorption processes. Due to the rapid uptake and short retention time  
 623 of radioactivity, the concentration of radionuclides in phytoplankton is calculated using  
 624 the equilibrium approach (equation 5):

$$C_{phyto} = CR_{phyto}C_w, \quad (7)$$

625 where  $CR_{phyto}$  (m<sup>3</sup>kg<sup>-1</sup>, wet weight) is the concentration ratio for phytoplankton.

626 Standard literature values for all these parameters for the four considered species may

	Zooplankton	Non-pisc. fish	Pisc. fish
$T_{0.5}$ (day)	5	75	200
$a$	0.2	0.5	0.7
$b$	0.001	0.001	0.001
$K_1$ (day <sup>-1</sup> )	1.0	0.035	0.0055
$K_w$ (m <sup>3</sup> /kg day)	1.5	0.1	0.075

Table 2: Parameters used in the dynamic BUM (from Maderich et al., 2014a). The concentration factor for phytoplankton is  $CR_{phyto} = 20$  l/kg.

627 be seen in Maderich et al. (2014a), which have been used in calculations. They are  
628 provided in Table 2.

### 629 **B.3 Allometry method**

630 An allometric method (Schmidt-Nielsen, 1977; West et al., 1997) may be used to simulate  
631 the radionuclide concentrations in marine biota. There is a variety of marine biota, with  
632 different sizes and shapes, in the oceans. Regardless of the variety of marine species,  
633 physical correlations between individual characteristics and masses have been provided.  
634 There are several allometric equations related to many biological characters including  
635 daily food ingestion rate, water intake rate and biological half-life of radionuclides. A  
636 first-order kinetics is used to predict radionuclide concentrations in marine biota based  
637 on an allometric equation (Schmidt-Nielsen, 1977; West et al., 1997). The general form  
638 of equation is as follows:

$$q = \frac{R}{k} (1 - e^{-kt}) \quad (8)$$

639 where  $q$  is the total activity (Bq) in the organism of concern at time  $t$ ,  $R$  is activity intake  
640 rate (Bq/day) into the organism,  $k$  is the effective loss rate of activity (1/day) from the  
641 organism, and  $t$  is the total length of exposure to the contaminant (day). Whole-body  
642 radionuclide concentration (Bq/kg) in the organism is calculated as  $q$  divided by the mass

643  $M$  (kg):

$$C_{biota} = \frac{R}{kM} (1 - e^{-kt}) \quad (9)$$

644 where  $C_{biota}$  is the radionuclide concentration in marine biota (Bq/kg).  $k$  can be expressed  
645 as follows:

$$k = \frac{\ln 2}{T_{1/2}} \quad (10)$$

646 The biological half-life of Cs for fish is (Higley et al., 2003):

$$T_{1/2} = 3.5 \times (1000 \times M)^{0.24} \quad (11)$$

647 In the case of phytoplankton and zooplankton constant values are used (Vives i Batlle et  
648 al., 2008; Table 1).

649 The activity intake rate is calculated as:

$$R = A_E F_I C_f + W_I C_w - E_R C_{biota} \quad (12)$$

650 where  $A_E$  is assimilation efficiency,  $F_I$  is food intake rate,  $C_f$  is  $^{137}\text{Cs}$  concentration in  
651 food,  $W_I$  is water intake rate and  $E_R$  is excretion rate (Brown et al., 2004; tables 5 and  
652 6).

653 Values for mass  $M$  of phytoplankton and zooplankton are taken from Vives i Batlle  
654 et al. (2008). For non-piscivorous and piscivorous fish they are respectively assumed as  
655 0.2 kg and 2 kg.

656 The semi-dynamic allometric method was applied to calculate radionuclide concen-  
657 trations in marine biota from radionuclide concentrations in seawater in LORAS model.  
658 This approach is a very efficient method, requiring short computational times (about one  
659 minute for a two year simulation in a Linux cluster).

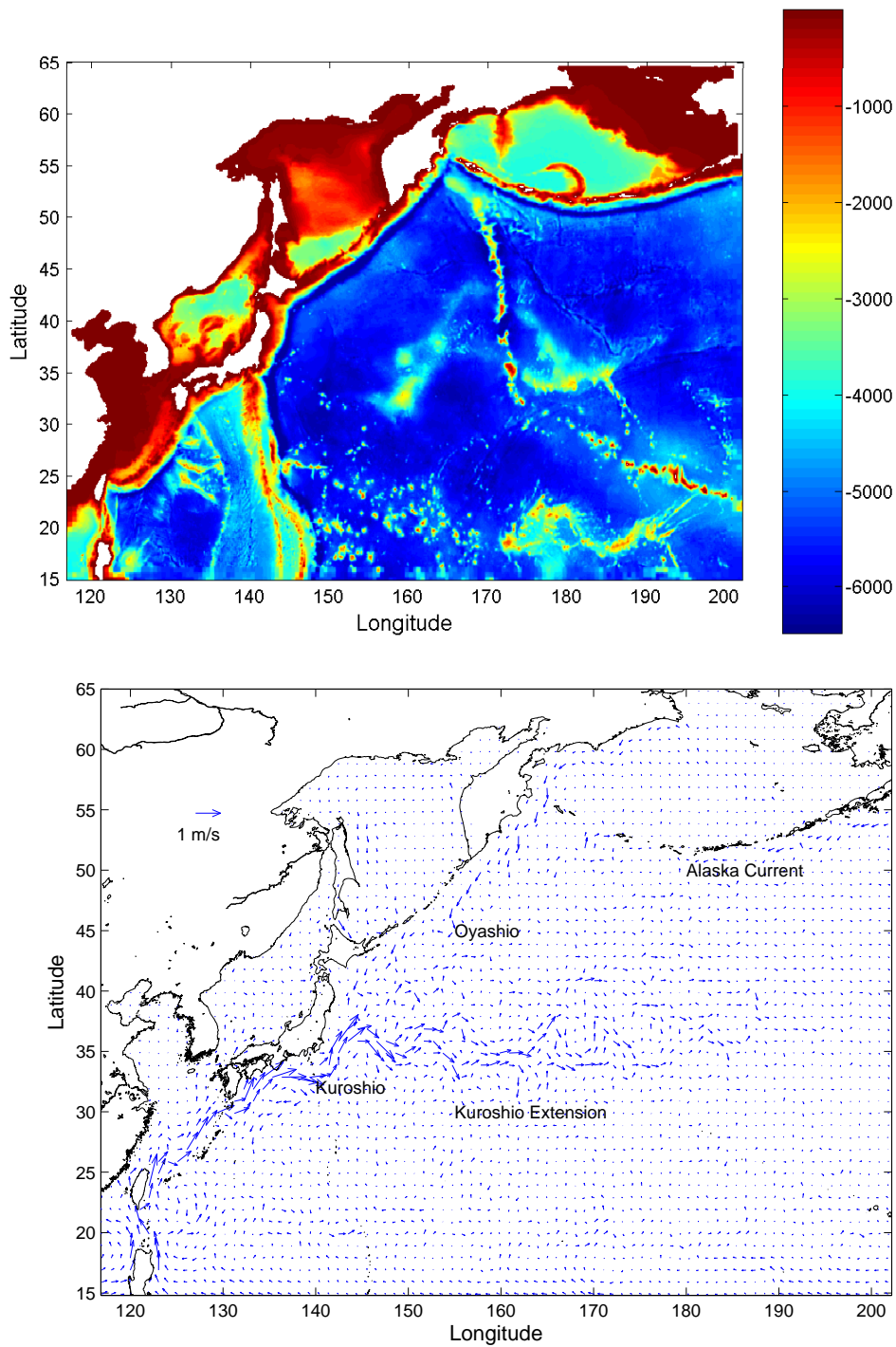


Figure 1: Water depths (m) over the model domain and average water circulation in March 2011, as an example, obtained with FORA model.

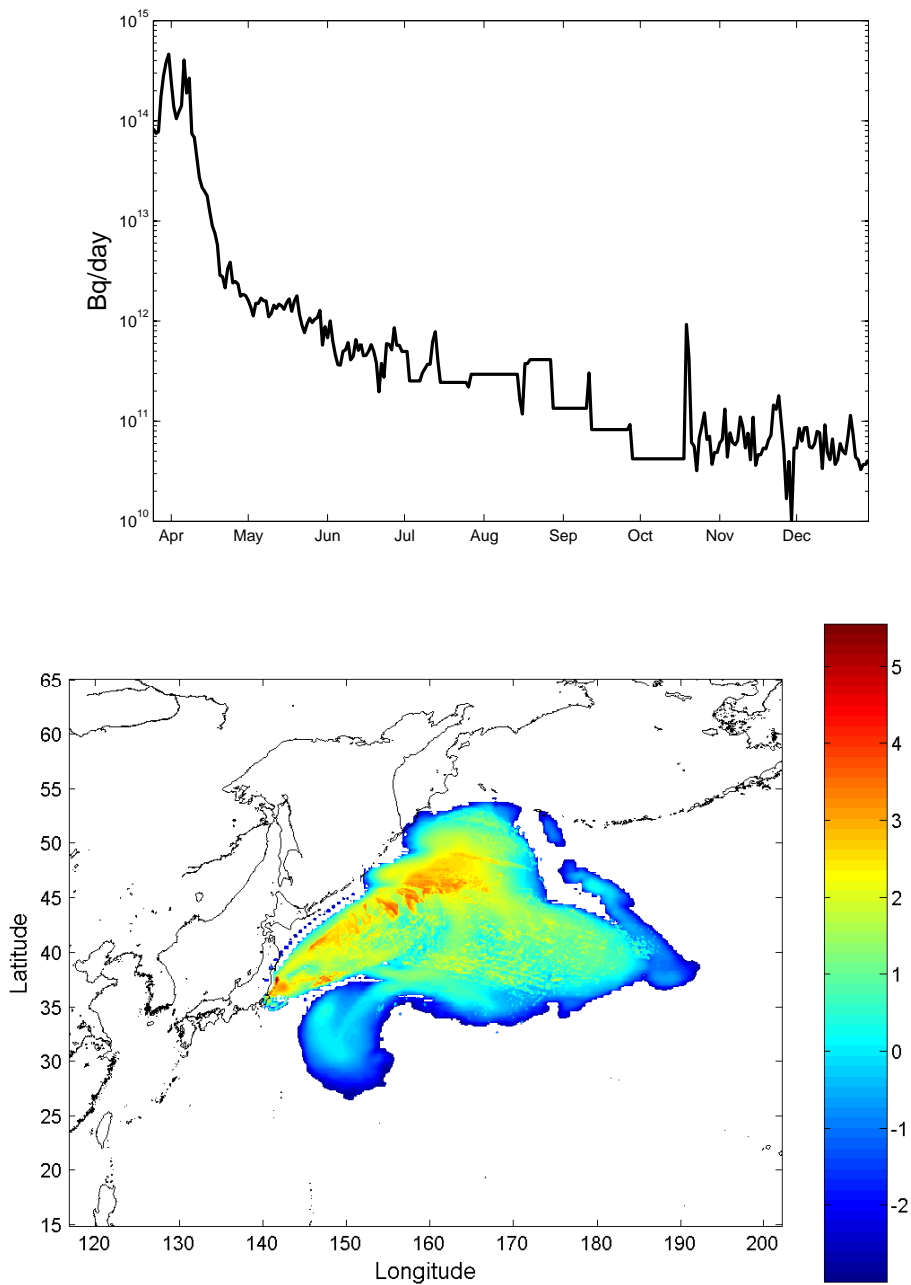


Figure 2: Top: Direct releases from FDNPP into the Pacific Ocean. Bottom: Example of atmospheric deposition for March 15th, 2011, in  $\text{Bq}/\text{m}^2$  (logarithmic scale).

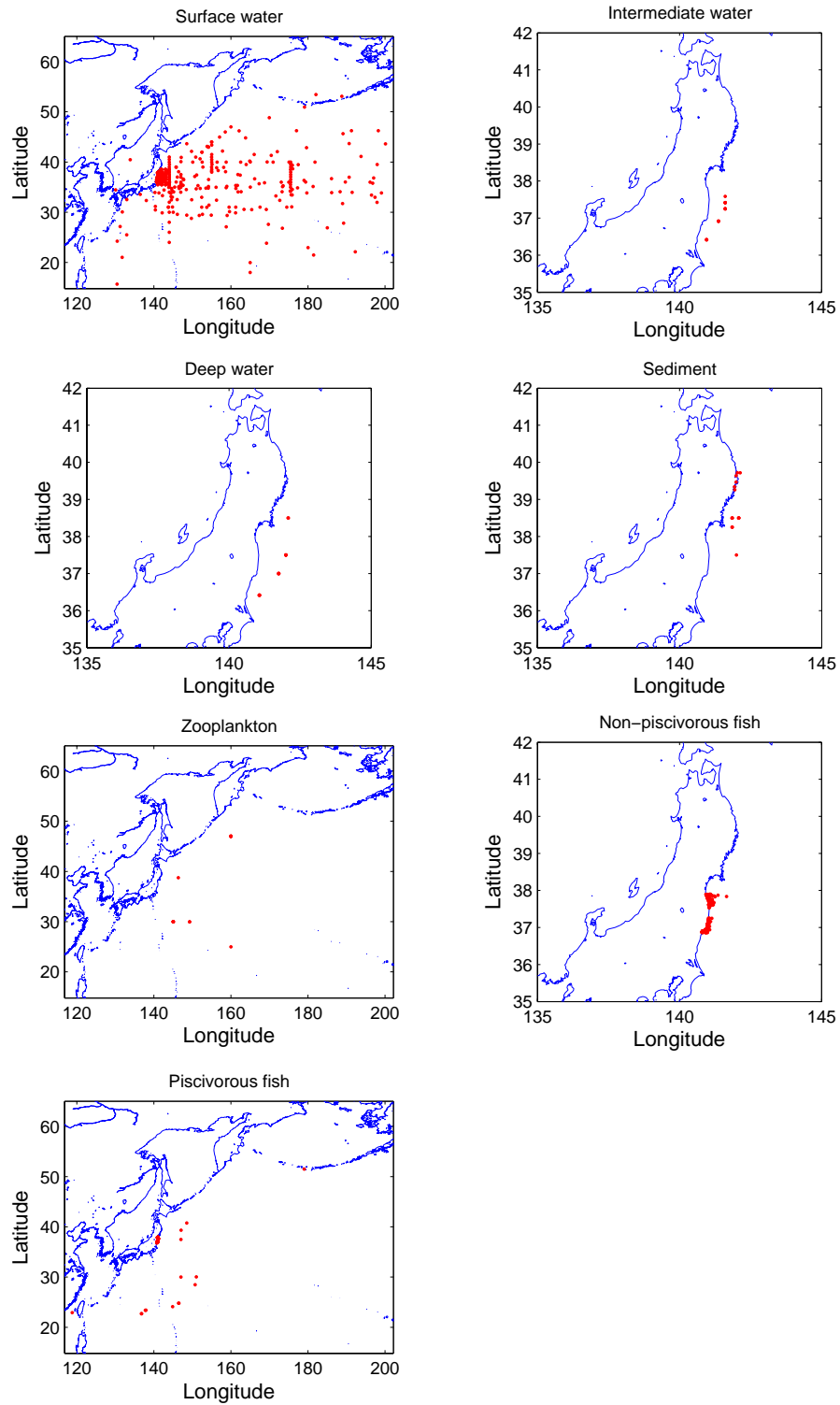


Figure 3: Locations of sampling points for all considered environmental compartments.



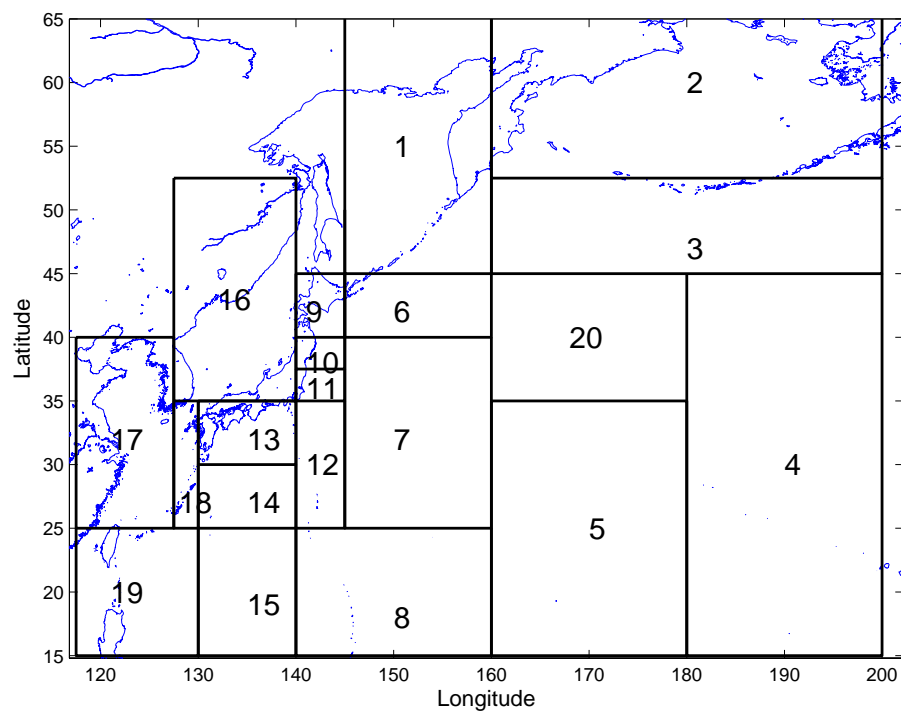


Figure 4: Box division of the Pacific Ocean for model/data comparisons.

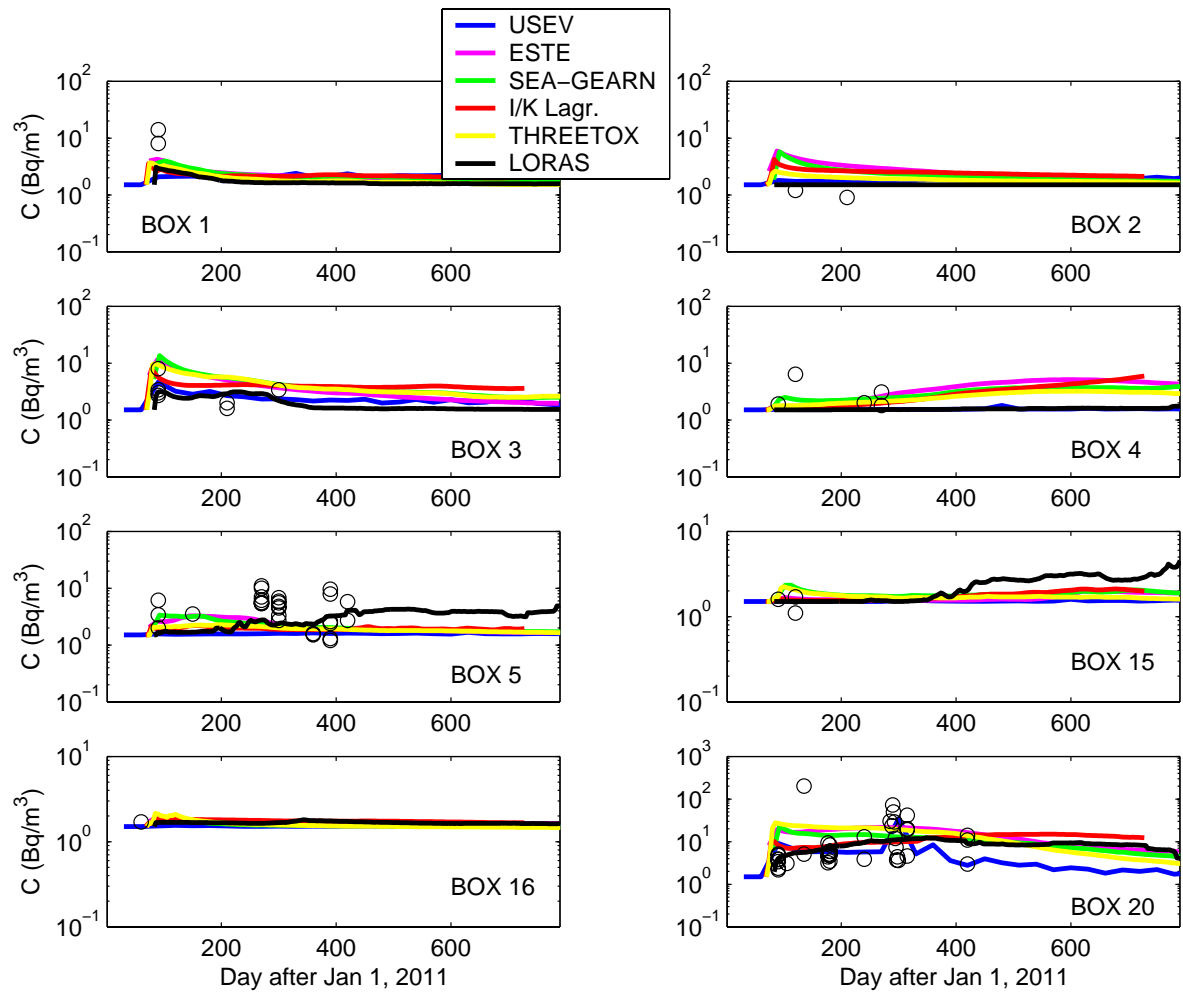


Figure 5: Model predictions and measurements of  $^{137}\text{Cs}$  concentrations in surface water for some boxes in the Pacific.

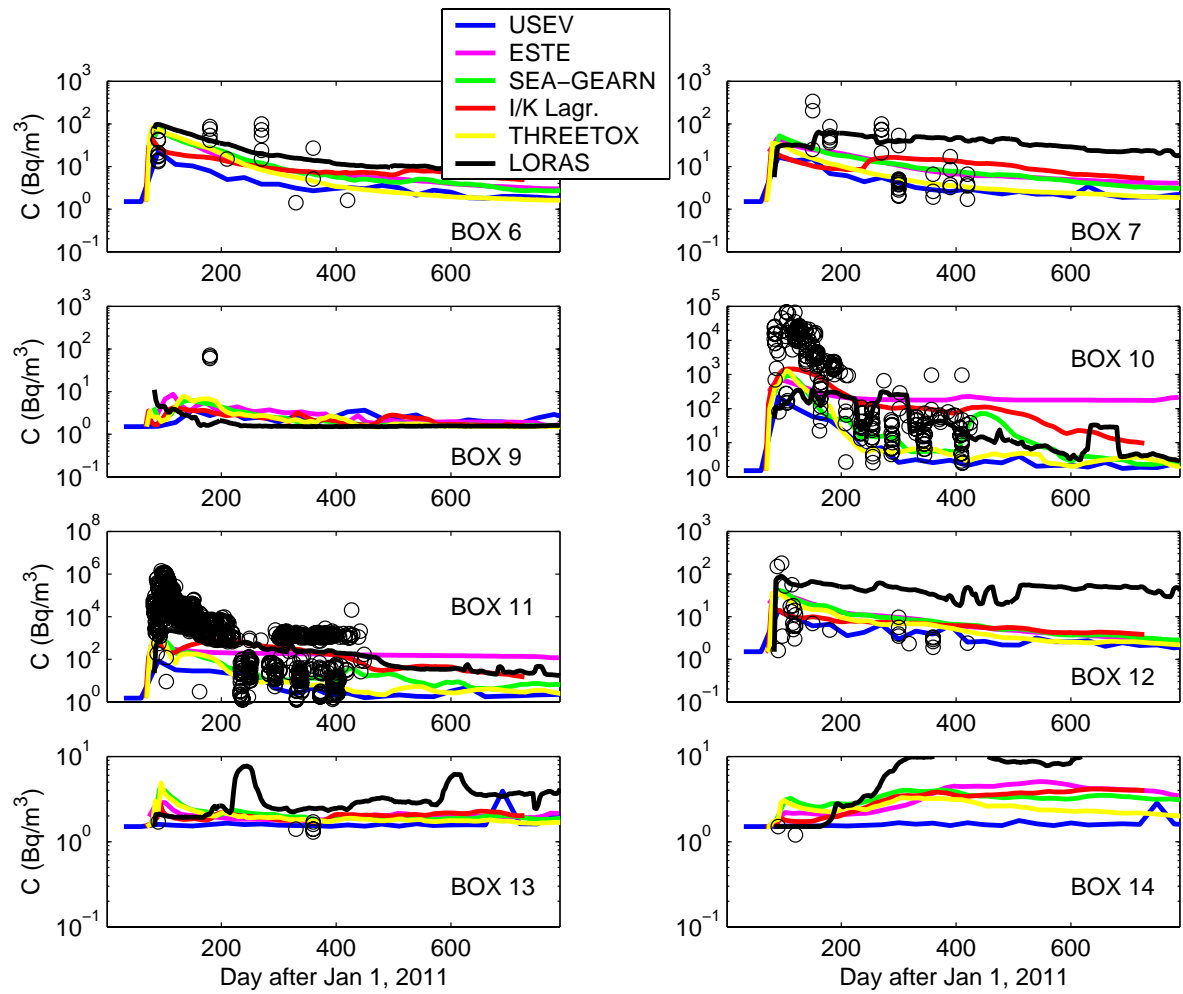


Figure 6: Model predictions and measurements of  $^{137}\text{Cs}$  concentrations in surface water for some boxes in the Pacific.

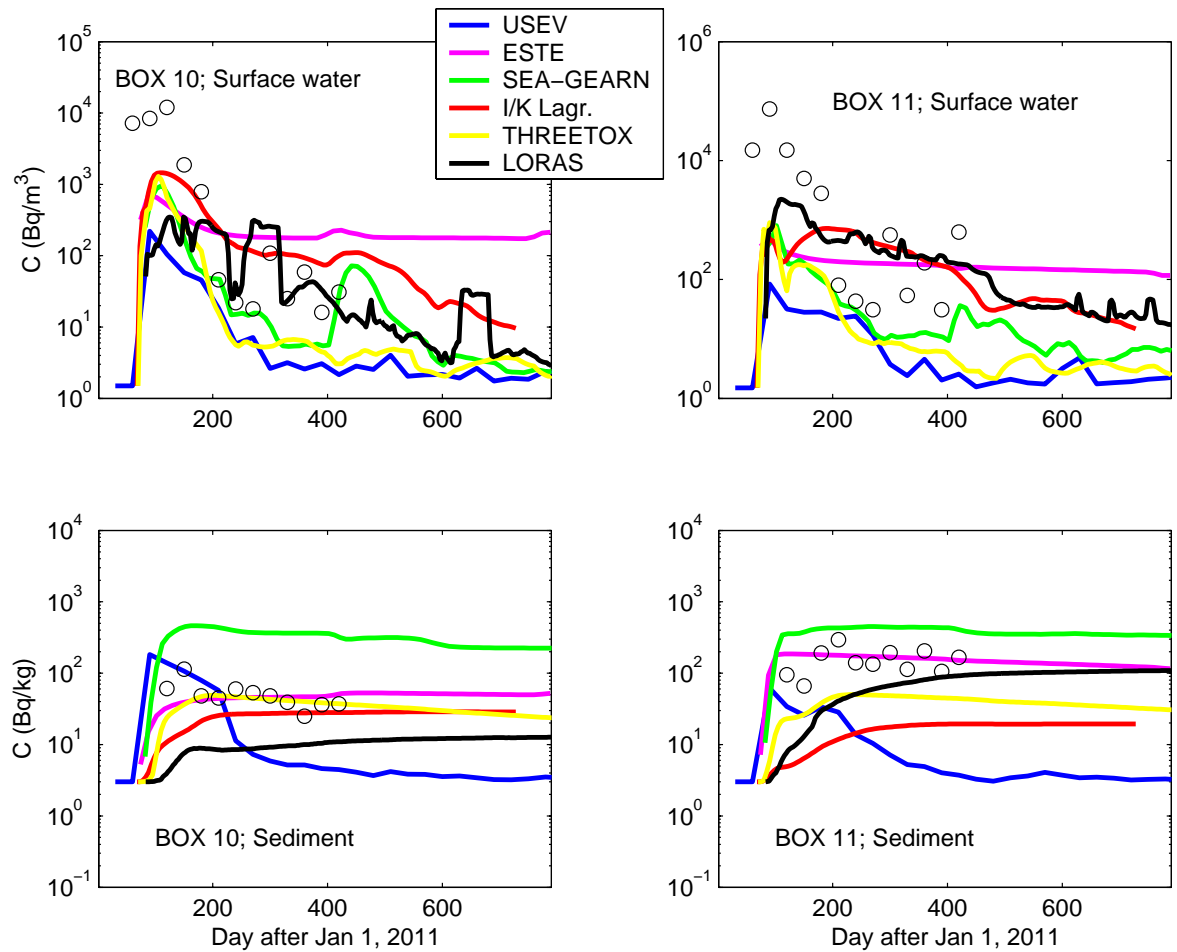


Figure 7: Model predictions and geometric means of  $^{137}\text{Cs}$  concentrations measured for each month in boxes 10 and 11, for surface water and sediments. Geometric standard deviations are not drawn because they are too small compared with the vertical scales.

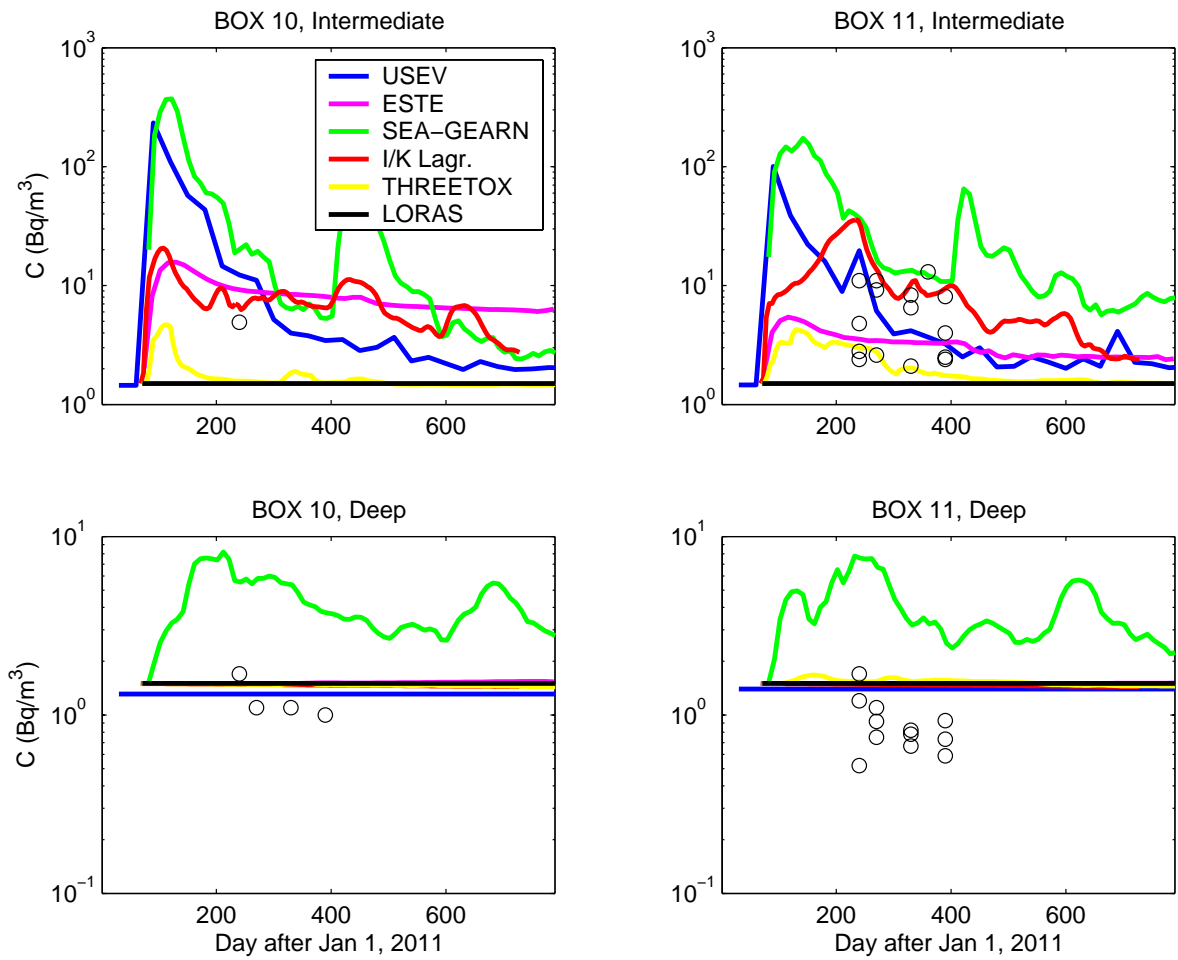


Figure 8: Model predictions and measurements of  $^{137}\text{Cs}$  concentrations in intermediate and deep waters for some boxes in the Pacific.

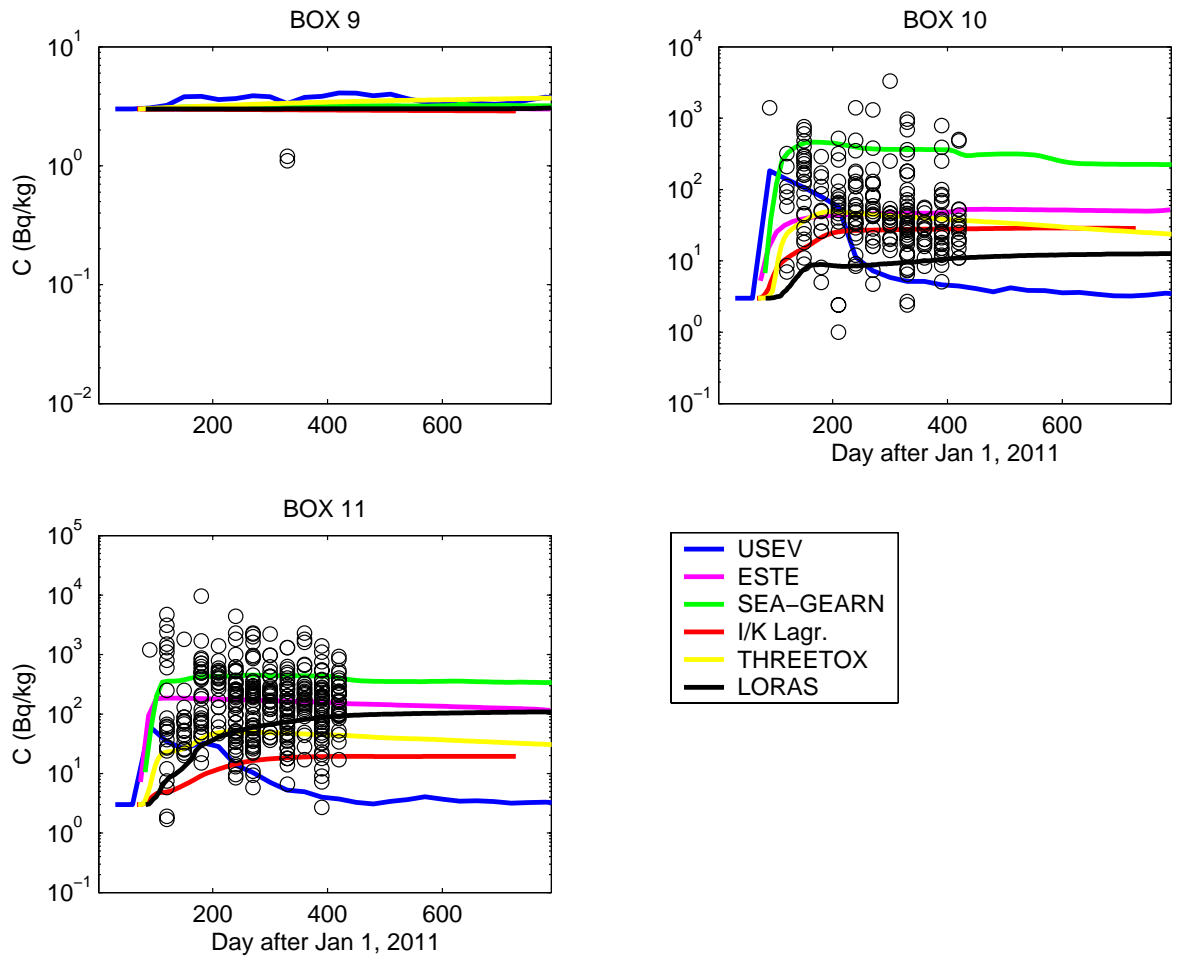


Figure 9: Model predictions and measurements of  $^{137}\text{Cs}$  concentrations in bed sediments for some boxes in the Pacific.

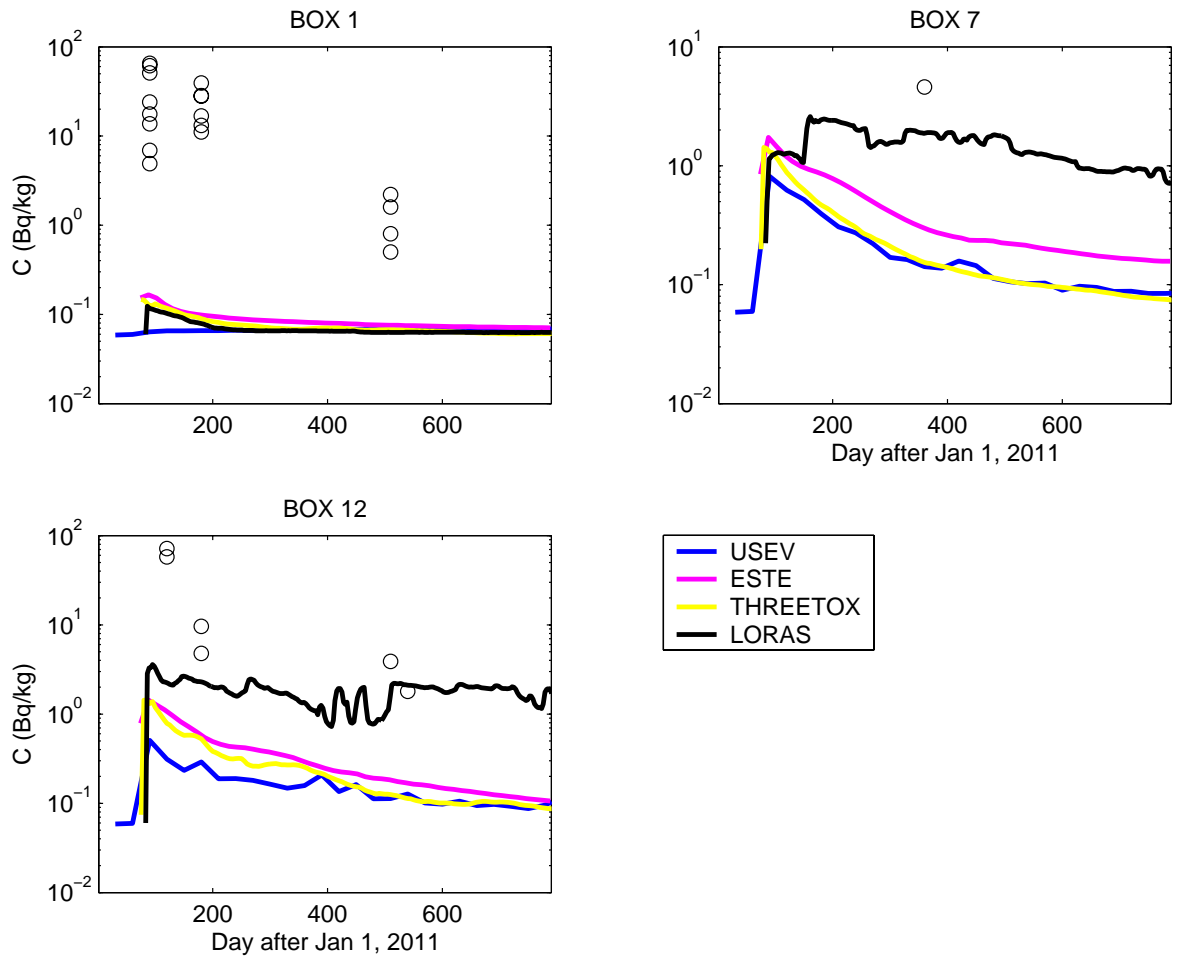


Figure 10: Model predictions and measurements of  $^{137}\text{Cs}$  concentrations in zooplankton for some boxes in the Pacific (Bq/kg wet weight).

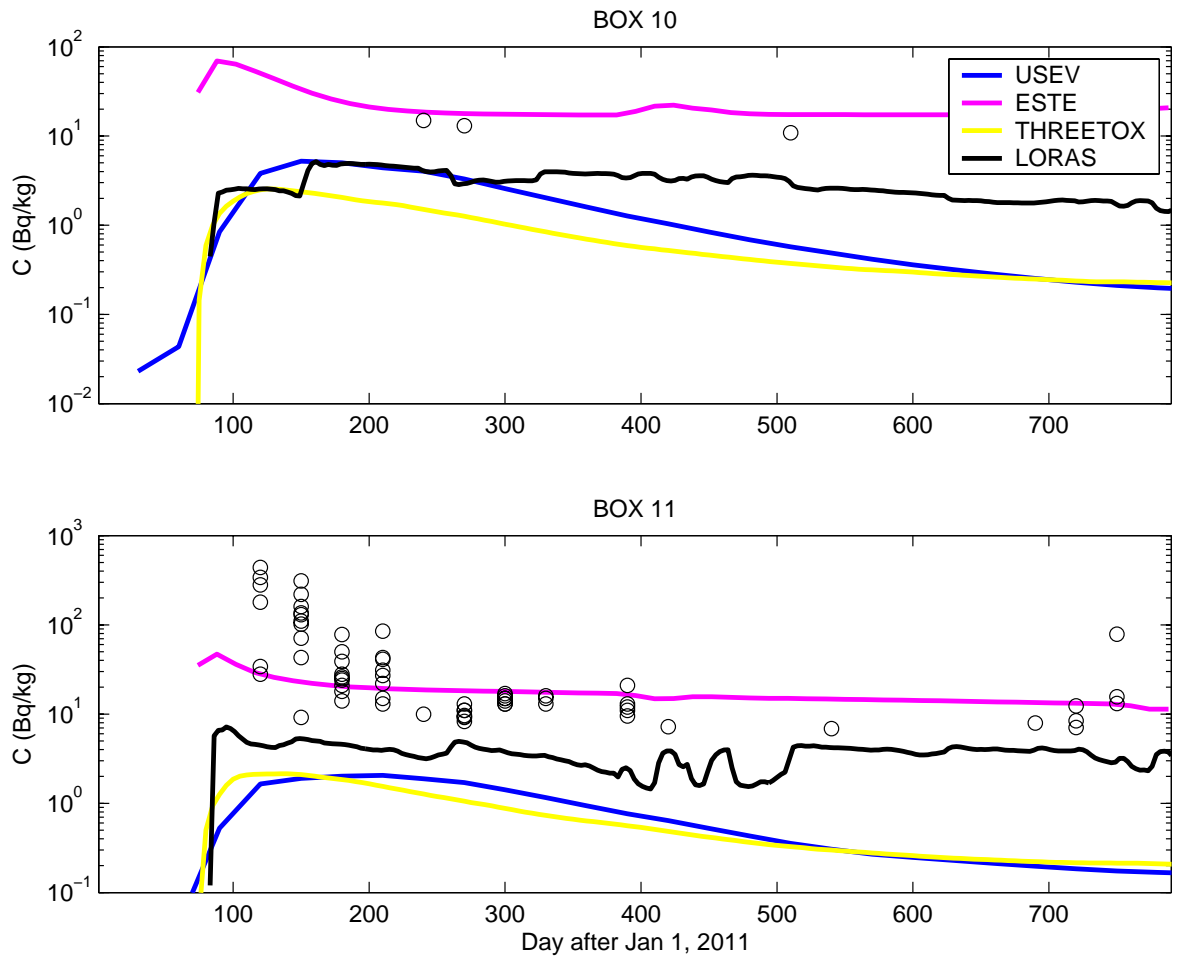


Figure 11: Model predictions and measurements of  $^{137}\text{Cs}$  concentrations in non-piscivorous fish (pelagic) for some boxes in the Pacific (Bq/kg wet weight).



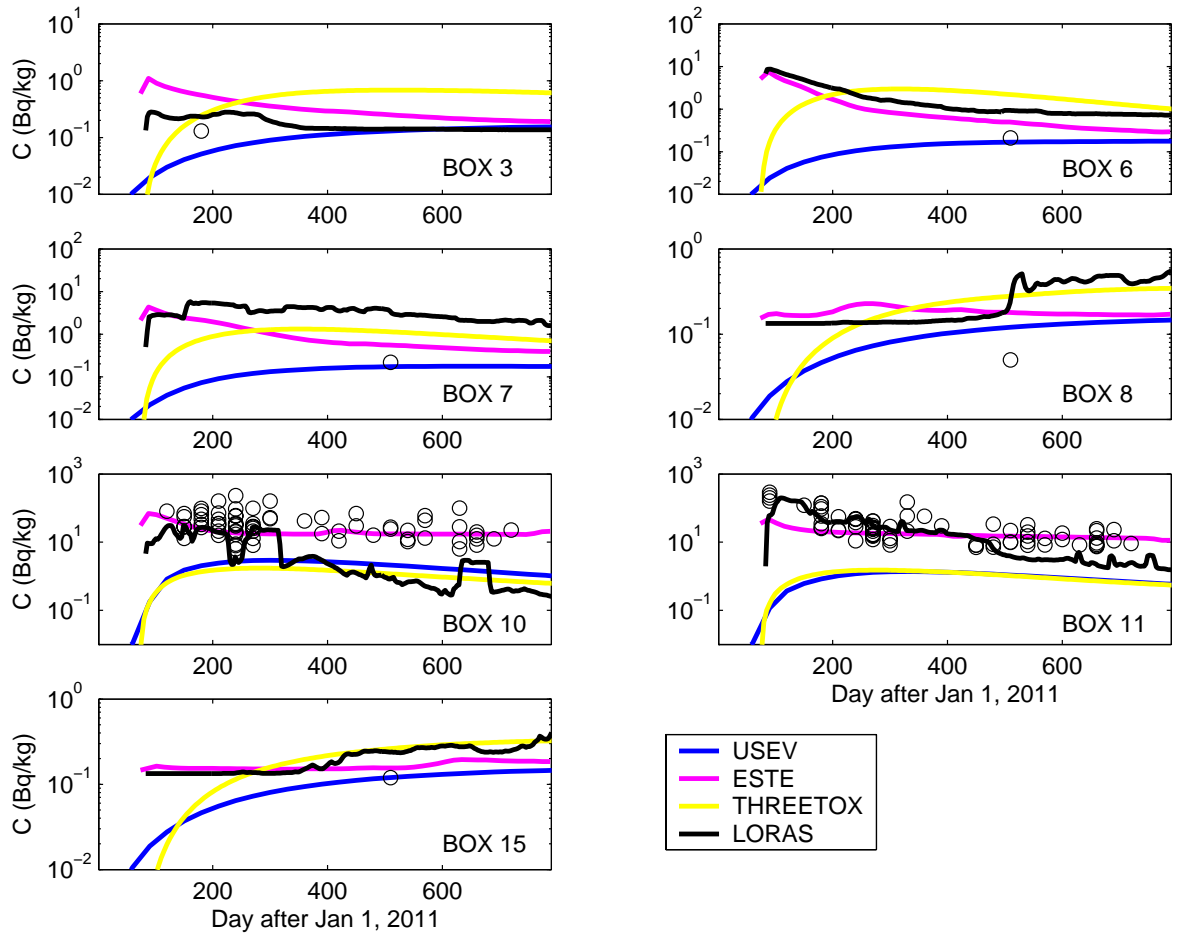


Figure 12: Model predictions and measurements of  $^{137}\text{Cs}$  concentrations in piscivorous fish (pelagic) for some boxes in the Pacific (Bq/kg wet weight).

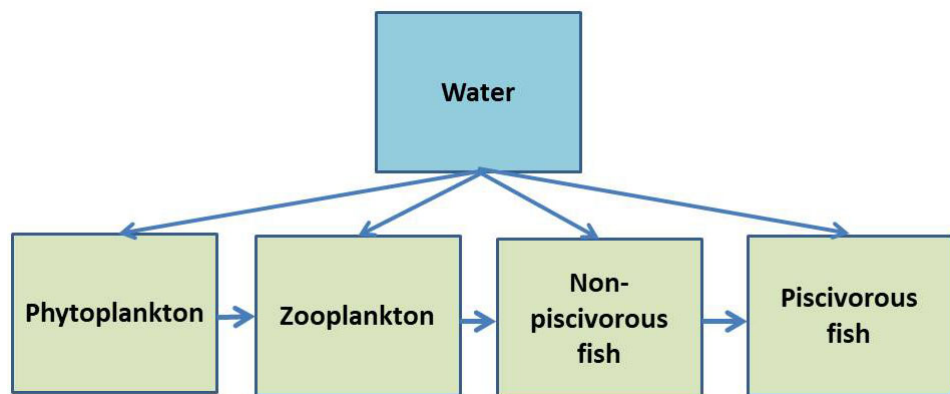


Figure 13: Scheme of radionuclide transfer (arrows) in a dynamic food chain model (from Maderich et al., 2014a).