

A multi-model assessment of the impact of currents, waves and wind in modelling surface drifters and oil spill

De Dominicis M.^{1,2,*}, Bruciaferri D.¹, Gerin R.³, Pinardi N.⁴, Poulain P.M.³, Garreau Pierre⁵, Zodiatis G.⁶, Perivoliotis L.⁷, Fazioli L.⁸, Sorgente R.⁸, Manganiello C.⁹

¹ Istituto Nazionale di Geofisica e Vulcanologia, Bologna, Italy

² National Oceanography Centre, Liverpool, United Kingdom

³ Istituto Nazionale di Oceanografia e di Geofisica Sperimentale (OGS), Trieste, Italy

⁴ Department of Physics and Astronomy, Alma Mater Studiorum, University of Bologna, Italy

⁵ IFREMER, Brest, France

⁶ Oceanography Centre, University of Cyprus (OC-UCY), Nicosia, Cyprus

⁷ Hellenic Center for Marine Research (HCMR), Athens, Greece

⁸ CNR- IAMC, Oristano, Italy

⁹ Italian Coast Guard Headquarters, Ministry of Infrastructures and Transports, Rome, Italy

* Corresponding author : M. De Dominicis, email address : micdom@noc.ac.uk

Abstract :

Validation of oil spill forecasting systems suffers from a lack of data due to the scarcity of oil slick in-situ and satellite observations. Drifters (surface drifting buoys) are often considered as proxy for oil spill to overcome this problem. However, they can have different designs and consequently behave in a different way at sea, making it not straightforward to use them for oil spill model validation purposes and to account for surface currents, waves and wind when modelling them. Stemming from the need to validate the MEDESS4MS (Mediterranean Decision Support System for Marine Safety) multi-model oil spill prediction system, which allows access to several ocean, wave and meteorological operational model forecasts, an exercise at sea was carried out to collect a consistent dataset of oil slick satellite observations, in-situ data and trajectories of different type of drifters. The exercise, called MEDESS4MS Serious Game 1 (SG1), took place in the Elba Island region (Western Mediterranean Sea) during May 2014. Satellite images covering the MEDESS4MS SG1 exercise area were acquired every day and, in the case an oil spill was observed from satellite, vessels of the Italian Coast Guard (ITCG) were sent in-situ to confirm the presence of the pollution. During the exercise one oil slick was found in-situ and drifters, with different water-following characteristics, were effectively deployed into the oil slick and then monitored in the following days. Although it was not possible to compare the oil slick and drifter trajectories due to a lack of satellite observations of the same oil slick in the following days, the oil slick observations in-situ and drifters trajectories were used to evaluate the quality of MEDESS4MS multi-model currents, waves and winds by using the MEDSLIK-II oil spill model. The response of the drifters to surface ocean currents, different Stokes drift parameterizations and wind drag has been examined. We found that the surface ocean currents mainly drive the transport of completely submerged drifters. The accuracy of the simulations increases with higher resolution currents and with addition of the Stokes drift, which is better estimated when provided by wave models. The wind drag improves the

modelling of drifter trajectories only in the case of partially emerged drifters, otherwise it leads to an incorrect reproduction of the drifters' direction, which is particularly evident in high speed wind conditions.

Keywords : Oil spill modelling, Drifters, Oil slick, Mediterranean, Met-ocean models

1. Introduction

Verification of oil spill forecasting capabilities is both a crucial issue and a difficult task to perform. The main reason for this is the lack of time series of oil slick observations, due to the long revisit time for satellites and the scarcity of in-situ data collected. The main datasets of remote sensing oil slick observations for oil spill validation were collected during the recent major accidental oil spills (*Prestige* Spain, 2002; Lebanon accident, 2006; *Deepwater Horizon*, Gulf of Mexico 2011) and were used for the evaluation of oil spill models forecasting accuracy (Carracedo et al., 2006; Coppini et al., 2011; Mariano et al., 2011; Liu et al., 2011). Ad-hoc oceanographic surveys can be also organized to collect in-situ observations of slicks for oil spill forecasting validation (Pisano, 2016) by combining real-time satellite observations with visual and instrumental inspection of the slicks. However, when real oil slick data are lacking, drifters might help for oil spill model validation. Drifters are oceanographic instruments used to study the surface circulation and oceanographic dynamics, they are designed to be transported by ocean currents and these peculiarities make them useful tools for the validation of models of Lagrangian particle dispersion (Reed et al. 1994; Al-Rabeh et al. 2000; Price et al. 2006; Caballero et al. 2008; Brostrom et al. 2008; Sotillo et al. 2008; Abascal et al. 2009; Zodiatis et al. 2010; Cucco et al. 2012; Sayol et al. 2014). Nowadays, several different kinds of drifters exist with different shape, size and immersion depth. When using drifters for oil spill model validation it is necessary to know which are the processes that affects the dynamics of different type of drifters. Indeed, each type of drifter behaves in a different way at sea and this should be carefully considered when using them as a proxy for oil spill. Furthermore, no study so far has been done to evaluate which type of drifter really follows an oil slick. The latter is extremely complicated at sea, because it requires the deployment of different type of drifters into a real oil slick, the observation of the oil slick evolution by subsequent satellite images or by aerial survey, together with the acquisition of the drifters' trajectories. Moreover, oil slick behavior may depend on oil quantity at sea that can be just a thin film at the surface or widely dispersed in the water column, making it even more complicated to find the ideal drifter representing an oil spill.

In the framework of the MEDESS4MS (Mediterranean Decision Support System for Marine Safety) project, which has been dedicated to the maritime risks prevention and strengthening of maritime safety related to oil spill pollution in the Mediterranean, a multi-model oil spill prediction service has been built using different oil spill numerical models and national ocean and meteorological forecasting systems, in order to deliver an operational multi-model oil spill prediction service for the entire Mediterranean Sea (Zodiatis et al., 2016). Stemming from the need to validate and evaluate the accuracy of the oil spill forecasts provided by the MEDESS4MS multi-model oil spill forecasting system, an exercise at sea, called MEDESS4MS Serious Game 1 (SG1), took place in the Elba Island region, Western Mediterranean (Figure 1), during May 2014 (17 - 27 May 2014). Satellite images covering the MEDESS4MS SG1 exercise area were acquired every day and, in the case an oil spill was observed from satellite, vessels of the Italian Coast Guard (ITCG) were sent in-situ to confirm the presence of the pollution. The ITCG vessels

*Corresponding author

Email address: micdom@noc.ac.uk (M. De Dominicis)

40 'ready to go' were located at the harbourmaster in Portoferraio and were equipped with drifters,
41 with different water-following characteristics, to be deployed into the oil slicks.

42 During the 10 days exercise two oil slick alerts were received from the satellite systems
43 monitoring the area (on 17 May 2014 and on 21 May 2014). One oil slick was found in-situ
44 and drifters were effectively deployed into the oil and then monitored during the following days.
45 Unfortunately, we did not succeed in the collection of a time series of observations of the same oil
46 slick by satellite over the same time, which would have been helpful for the comparison between
47 oil slick and drifter behavior. Using the drifter data collected, the main objectives of this paper
48 are: (i) evaluation of the quality of MEDESS4MS multi-model currents, waves and winds; (ii)
49 comprehension of the differences between different drifter behavior at sea and assessment of the
50 capabilities to simulate them.

51 The paper is organized as follows: Section 2 describes the data collected for oil spill forecast-
52 ing validation (remote sensing data, in-situ data and drifters); Section 3 presents the modelling
53 methodology used and the description of the experiments performed; Section 4 reports the vali-
54 dation results, and Section 5 summarizes our conclusions.

55 2. Data

56 2.1. Data from satellite

57 The MEDESS4MS SG1 aimed to detect oil slicks by satellite using SAR images covering the
58 exercise area available through CleanSeaNet2 (CSN-2) and COSMOSKYMED (CSK) services
59 for the entire exercise period. The acquisition of two satellite images every day from CSN-2 or
60 from CSK were planned. The CSK system allows a more frequent revisit time of the same area
61 (12 hours, depending on the size of the area). However, the planned images were not available
62 every day.

63 During the exercise period, two oil slick alerts were received from the satellite monitoring
64 systems. The first one was on the morning of the 17th May 2014 at 05:38 UTC, when an oil spill
65 was observed by CleanSeaNet2 (CSN-2). Figure 2-a shows the original satellite image, while
66 Figure 2-b is the output of the CSN-2 automatic detection algorithm (dark area in the northern
67 part of the domain of Figure 2-a represents a low wind area or ocean features, not an oil slick).
68 The oil spill was reported as being composed of 7 oil slick patches, the centre positions of those
69 are reported in Table 1. The ITCG vessel was sent to confirm the oil spill in-situ, which was
70 found at 07:08 UTC by visual detection (i.e. iridescence). The oil was sampled in 4 different
71 positions reported in Table 2.

72 The second slick alert was received on the morning of the 21st of May 2014 at 05:07 UTC,
73 when two oil slicks were detected by the CSK satellite system, west of Elba Island (see Figure 1).
74 The ITCG vessel went immediately to search in-situ. However, by visual inspection of the area,
75 it was not possible to identify any oil slick. In addition, an ITCG plane ATR 42 equipped with
76 a side-looking airborne radar (SLAR) surveyed the area, and it confirmed that it was not possible
77 to detect any oil slick.

Slick ref.	Latitude	Longitude	Area (km ²)
A	42°57.93' N	10°00.42' E	0.97
B	42°57.40' N	09°58.98' E	0.03
C	42°57.63' N	09°59.53' E	0.16
D	42°57.36' N	10°01.00' E	0.24
E	42°58.70' N	10°01.27' E	0.09
F	42°57.17' N	09°58.68' E	0.16
G	42°58.55' N	10°00.93' E	0.13

Table 1: List of slicks composing the spill observed on the 17 May 2014 at 05:38 UTC by the CSN-2 system.

SAMPLE ID	Time (UTC)	Latitude	Longitude
M1	7:08	42°58.01' N	9°59.26' E
M2	7:35	42°58.82' N	9°59.34' E
M3	8:27	42°57.90' N	9°58.33' E
M4	8:49	42°58.21' N	9°59.37' E

Table 2: Oil Sampling positions and time on the 17 May 2014.

78 2.2. Drifters

79 Drifters were released into the observed oil slick on the 17th May 2014. In particular, in
80 order to be able to distinguish between the uppermost meter of the water column and the purely
81 superficial flow, some drifters with different water-following characteristics (CODE, iSLDMB,
82 iSPHERE, MAR-GE/T) were released inside the oil slick.

83 CODE surface drifters (Davis, 1985) are made of a 1 m long vertical tube with four wings
84 extending radially from the tube over its entire length. When in water they are completely sub-
85 merged, except for a small antenna on the top of the tube and four small floats attached on the
86 upper extremities of the wings. The design of iSLDMB (Iridium Self-locating Marker Buoy)
87 drifters is based on the CODE/Davis style oceanographic surface drifters, but they are made of
88 a 60 cm vertical tube. The CODE drifter is designed to minimize the effect of the wind on the
89 emerged part of the instrument (Poulain, 1999). iSPHERE (Iridium SPHERE) surface drifters
90 are 39.5 cm diameter spheres (Price et al., 2006), where in water the iSPHERE drifter is half
91 submerged. MAR-GE/T drifters have a cylindrical shape, with a diameter of 13.4 cm and height
92 of 28 cm, when in water the drifter is submerged for 1/3 of its height.

93 The deployment of the different types of drifters was aimed at revealing the proportions in
94 which the drifters follow the wind, currents and waves. The deployment time and positions are
95 listed in Table 3. The 2 iSPHERE drifters and the 2 CODE drifters were recovered after 1 day at
96 sea, while the iSLDMB and MAR-GE/T were recovered after 7 days at sea. From Fig. 3-a it is
97 possible to observe that CODE and iSLDMB followed a similar trajectory, while they moved in a
98 different direction and slower than the iSPHERE and MAR-GE/T drifters. It is worth noting that
99 2 CODE drifters were moving together, the same can be observed for the 2 iSPHERE drifters.
100 This allow us to be more confident in saying that the different behavior is due to the different
101 drifter shapes, rather than to sub-mesoscale ocean processes. From Fig. 3-b it is evident that,

102 after a few hours at sea, iSLDMB and MAR-GE/T diverged and described a completely different
 103 trajectory.

Drifter Type	Time (UTC)	Latitude	Longitude
iSPHERE	11:38	42°58.23'N	9°59.03' E
iSPHERE	11:39	42°58.18'N	9°59.03' E
MAR-GE/T	11:39	42°58.14'N	9°59.01' E
CODE	12:08	42°58.29'N	9°58.83' E
iSLDMB	12:00	42°58.17'N	9°58.86' E
CODE	12:04	42°58.16'N	9°58.79' E

Table 3: Positions and time of drifter deployments into the oil slick on 17 May 2014.

104 3. Method

105 Lagrangian models are a powerful tool for modelling the marine transport of pollution or
 106 floating objects. A Lagrangian model tracks a set of tracer particles forward in time from a
 107 source. In this work, particles can represent a floating object or surface oil slick and they can be
 108 passively transported by the upper ocean currents, waves and winds. In the case of an oil slick
 109 or any other pollutant, the Lagrangian model should also reproduce the transformation processes
 110 that affect the pollutant physical and chemical characteristics. In this study our attempt is to
 111 differentiate the effects of the currents and waves on the transport of a surface oil slick or floating
 112 object and to correctly simulate the effect of the wind on different kind of drifters.

113 The first objective of this work is to evaluate the quality of the multi-model forcing data
 114 (currents, wave, winds) provided by the MEDESS4MS system, that can drive the oil slick or
 115 floating object motion at sea. MEDESS4MS allows access to the forecasts of temperature and sea
 116 currents from different ocean models with variable horizontal spatial resolution and with 1-hour
 117 temporal resolution, as well as access to wave conditions and winds in a variety of temporal and
 118 spatial resolution (full description of the MEDESS4MS ocean, wave and meteorological model
 119 is given in Zodiatis et al. (2016)). The MEDESS4MS multi-model oil spill prediction system
 120 allows the use of 4 different oil spill models to forecast the trajectory of the oil slick observed by
 121 the satellite system and of the drifters released in the area of the experiment. However, only one
 122 of the MEDESS4MS oil spill models, MEDSLIK-II (see Sect. 3.1), has been used in this work to
 123 simulate the drifters' trajectories and the oil slick observed from satellite. This choice has been
 124 done to reduce the degrees of freedom, as it is necessary to evaluate the quality of the different
 125 ocean, wave and wind forcings. All the MEDESS4MS models that can provide sea, wave and
 126 wind state for the SG1 region have been used. A full description of the combination of models
 127 used is given in Sect. 3.2.

128 The second objective of this study is to understand the magnitude of the wave-induced trans-
 129 port. The tracer transport is mainly generated by the upper ocean currents and when dealing
 130 with models we must take care to understand what are the upper ocean currents provided by
 131 the hydrodynamic models and which processes are represented or not by the model data. When
 132 talking about advection by upper ocean currents we must specify that this includes parts which
 133 are due to the mean drift due to surface waves (Stokes, 1847) and mean currents forced by the
 134 waves. The Stokes drift is not represented by a hydrodynamic Eulerian model, while the mean

135 currents forced by the waves can be represented by an hydrodynamic model only when it is cou-
 136 pled with a wave model. In the present work the hydrodynamic models used (available from
 137 the MEDESS4MS system) do not include the interaction between wave momentum and current
 138 momentum. Although the interaction between the Ekman current and Stokes drift can be even
 139 more important than the Stokes drift itself (Smith, 2006; McWilliams et al., 2004; Mellor, 2003,
 140 2008; Ardhuin et al., 2008), the exact representation of the wave effect on the mean current field
 141 is still being debated and this term it is still not included in state-of-the-art hydrodynamic models.
 142 Thus, in this work, we cannot examine the effect of wave-induced currents on tracer transport,
 143 but we explore and compare different approaches to compute the Stokes drift, focusing on the
 144 benefit of having a dedicated forecasting wave system that can provide wave statistics to be fed
 145 into the oil spill/trajectory model.

146 The third objective is to shed light on how to correctly use the wind velocity in the simulation
 147 of tracers or drifters transport. In the past, the drift velocity of a surface oil slick or surface drifter
 148 was considered to be the sum of a fraction of the wind velocity (wind drift or wind correction)
 149 and an estimate of the current fields from OGCM. The wind correction has been interpreted in
 150 different ways in the past thirty years. Initially, it was necessary in order to reproduce the surface
 151 Ekman currents, i.e., the local wind effects that were not properly resolved by low-resolution,
 152 climatological models. A practical 'rule of thumb' was to add to the ocean currents a wind drift
 153 assumed to be 3% of the wind velocity and with a deviation angle between 0° and 25° (Al-Rabeh,
 154 1994; Reed et al., 1994). Nowadays, a correct representation of the Ekman currents is provided
 155 by OGCMs and that kind of wind correction is now obsolete. In more recent works, a 1% of the
 156 wind velocity has been added to the ocean currents in the same direction of the wind (0° deviation
 157 angle) to roughly represent the effect of waves, as done by Coppini et al. (2011) and Liubartseva
 158 et al. (2015). Indeed, as demonstrated by De Dominicis et al. (2013a) that approach is almost
 159 equivalent to the calculation of the Stokes drift using the empirical JONSWAP wave spectrum
 160 parameterization. Thus, if the upper ocean currents are correctly reproduced by an hydrodynamic
 161 model and a wave forecasting system is available, there is no need to add a wind correction to
 162 the ocean surface currents. Although these are both a fraction of the wind velocity, the above
 163 described wind corrections must not be confused with the direct drag of the wind on a floating
 164 objects that can be modelled by assessing the leeway (Breivik et al., 2011). In this work, by using
 165 drifters with different immersion depth we want to show how the wind drag effect (leeway) has
 166 a substantial importance only in the case of partially emerged floating objects.

167 3.1. MEDSLIK-II

168 The oil spill model code MEDSLIK-II (De Dominicis et al., 2013b,a; Bruciaferri and MEDSLIK-
 169 II system team, 2015) is designed to be used to predict the transport and weathering of an oil spill
 170 or to simulate the movement of a floating object. MEDSLIK-II is a Lagrangian model, which
 171 means that the oil slick is represented by a number N of constituent particles that move by ad-
 172 vection and disperse by Lagrangian turbulent diffusion. The advection is taken to be the sum of
 173 different components:

$$174 \begin{aligned} d\mathbf{x}_k(t) = & [\mathbf{U}_C(x_k, y_k, t) + \mathbf{U}_S(x_k, y_k, t) + \mathbf{U}_W(x_k, y_k, t) \\ & + \mathbf{U}_D(x_k, y_k, t)] dt + d\mathbf{x}'_k(t) \end{aligned} \quad (1)$$

174 where \mathbf{U}_C is the wind, buoyancy and pressure driven large scale current velocity field, \mathbf{U}_W is the
 175 wind-driven sea surface currents velocity correction term, \mathbf{U}_S is the wave-induced current term

176 (Stokes drift velocity), U_D is the wind drag velocity due to emergent part of the objects at the
177 surface and $d\mathbf{x}'_k(t)$ is the displacement due to the turbulent diffusion.

178 The term U_C represents the surface currents that can be provided by an oceanographic model.
179 However, in numerical circulation models the surface velocity represents the mean velocity in the
180 surface layer that can vary from few centimeters to few meters depending on the vertical model
181 discretization. Thus, surface currents from an oceanographic model do not actually represent the
182 currents at 0 m, but are just the currents at the first level of the model.

183 Several approaches exist to account for the term U_S ; the Stokes drift can be approximated
184 using only wind speed and direction, and it can be written as:

$$\begin{aligned} U_S &= D_S \cos \vartheta \\ V_S &= D_S \sin \vartheta \end{aligned} \quad (2)$$

185 where (W_x, W_y) are the wind velocity components at 10 m, $\vartheta = \arctg\left(\frac{W_y}{W_x}\right)$ is the wind direc-
186 tion and D_S is the Stokes drift velocity intensity in the direction of the wave propagation, at the
187 surface and for deep-water waves, is defined as:

$$D_S(z=0) = 2 \int_0^{\infty} \omega k(\omega) S(\omega) d\omega \quad (3)$$

188 where ω is angular frequency, k is wave-number, and $S(\omega)$ is wave spectrum. The wave
189 spectrum, $S(\omega)$, can be calculated using empirical parameterizations. MEDSLIK-II allows to
190 use Joint North Sea Wave Project (JONSWAP) spectrum parameterization (Hasselmann et al.,
191 1973), that describes the wave spectrum as a function of wind speed and fetch. Using this
192 parameterization a separate wave model is not needed, however, we assume that wind and waves
193 are aligned and the waves are generated only by the local wind, something that is not always the
194 reality (swell waves are not considered). This approach has been used in the past (De Dominicis
195 et al., 2013a, 2014) showing that the addition of this term is almost equivalent to addition of 1%
196 of the wind velocity and leads to an improvement of the modelled trajectories. If wave model data
197 are available, the Stokes drift can be calculated from wave statistics that can be provided by any
198 wave model: the significant wave height, H_s , wave mean period, \bar{T} , and wave mean direction, $\bar{\phi}$.
199 This is a new MEDSLIK-II feature, recently added during the MEDESS4MS project (Bruciaferri
200 and MEDSLIK-II system team, 2015). The Stokes drift velocity intensity can be then calculated
201 by rewriting Eq. (3) as:

$$D_S(z=0) = \frac{1}{8} \bar{\omega} \bar{k} H_s^2 \quad (4)$$

202 where the significant wave height, H_s , is approximated as four times the square root of the
203 zeroth-order moment of the wave spectrum, the wave mean angular frequency is $\bar{\omega} = \frac{2\pi}{\bar{T}}$ and the
204 mean wave-number is $\bar{k} = \frac{\bar{\omega}^2}{\bar{T}}$ (for deep-water). Thus, the Stokes drift velocity components are:

$$\begin{aligned} U_S &= D_S \cos \bar{\phi} \\ V_S &= D_S \sin \bar{\phi} \end{aligned} \quad (5)$$

205 where $\bar{\phi}$ is the wave mean direction provided by wave models. The Stokes drift calculated
206 using bulk (statistically averaged) wave parameters, as in Eq. 4, can result in an underestimation
207 of the Stokes drift induced-current (Tamura et al., 2012). Alternatively, third-generation wave

208 models can directly give the surface Stokes drift velocity, i.e. integration of the full wave spec-
 209 trum done internally by the wave model. Unfortunately, the latter was not implemented by the
 210 wave models available in the MEDESS4MS system.

211 The local wind correction term \mathbf{U}_W is written as:

$$\begin{aligned} U_W &= \alpha(W_x \cos\beta + W_y \sin\beta) \\ V_W &= \alpha(-W_x \sin\beta + W_y \cos\beta) \end{aligned} \quad (6)$$

212 where W_x and W_y are the wind zonal and meridional components at 10 m respectively and
 213 α is the percentage of the wind to be considered in the oil slick transport and β is the angle of
 214 deviation with respect to the currents direction. When \mathbf{U}_C is provided by oceanographic models
 215 that resolve the upper ocean layer dynamics (with fine vertical resolution and using turbulence
 216 closure sub-models), the term \mathbf{U}_C contains a satisfactory representation of surface ageostrophic
 217 currents and the \mathbf{U}_W term may be neglected (in this work \mathbf{U}_W has been always set equal to 0).

218 The wind drag velocity, \mathbf{U}_D , is associated with the leeway (windage) of a floating object,
 219 defined as the drift associated with the wind force on the overwater structure of the object.
 220 As defined by Breivik et al. (2011) and Röhrs et al. (2012) the leeway-drift velocities can be
 221 parametrized as follows:

$$\mathbf{U}_D = \sqrt{\frac{\rho_a A_a C d_a}{\rho_w A_w C d_w}} \mathbf{W} = \gamma \mathbf{W} \quad (7)$$

222 where \mathbf{W} is the wind velocity at 10 m, ρ , A , C_d are the fluid density, projected areas of the
 223 object and drag coefficient, respectively, and subscripts a and w denote the air and seawater
 224 environments. The leeway factor γ cannot be calculated directly because the drag coefficients
 225 $C d_a$ and $C d_w$ are dependent on Reynolds numbers and are not straightforward to use at the air-sea
 226 interface with wave disturbances (Röhrs et al., 2012). However, if we assume $C d_a = C d_w = 1$
 227 (Richardson, 1997) and the density of air and water are considered to be $\rho_a = 1.29 \text{kgm}^{-3}$ and
 228 $\rho_w = 1025 \text{kgm}^{-3}$, in the particular case of the over-water structure and the submerged part of
 229 the object being the same, the parameter γ is equal to 0.035. However, as reported by Breivik
 230 et al. (2011) the choice of both drag coefficients to be equal to 1 might not take into account the
 231 heave, pitch and roll of open ocean conditions which induce additional viscous damping and drag
 232 coefficients are not straightforward to evaluate at the air-sea interface with wave disturbances
 233 (Röhrs et al., 2012). Field experiments using SPHERE drifters, performed by Röhrs et al. (2012),
 234 suggest to use γ in the range 0.003 - 0.01.

235 The last term of Eq. 1 is due to turbulent diffusion and it is parameterized with a random
 236 walk scheme as

$$d\mathbf{x}'_k(t) = \sqrt{2\mathbf{K}dt} \vec{Z} \quad (8)$$

237 where \mathbf{K} is the turbulent diffusion diagonal tensor and \vec{Z} is a vector of independent random
 238 numbers used to model the Brownian random walk processes chosen for the parametrization of
 239 turbulent diffusion. The turbulent diffusion is considered to be horizontally isotropic and the
 240 three diagonal components of \mathbf{K} are indicated by K_h, K_h, K_v . In the simulation experiments of a
 241 real oil slick, K_h has been set to $2 \text{m}^2\text{s}^{-1}$, in the range $1 - 100 \text{m}^2\text{s}^{-1}$ indicated by ASCE (1996)
 242 and De Dominicis et al. (2012), while K_v has been set to $0.01 \text{m}^2\text{s}^{-1}$ in the mixed layer (assumed
 243 to be 30 m deep) and below it to $0.0001 \text{m}^2\text{s}^{-1}$. When simulating drifter trajectories, K_h has been
 244 set to $2 \text{m}^2\text{s}^{-1}$, assuming that in the region there is not any feature that can break the isotropy

245 or change dramatically the horizontal diffusivity. The drifter trajectory is assumed to be the
246 barycentre of the particle cloud, while the vertical diffusivity coefficients are set to zero, as the
247 vertical movement of the drifter is not allowed.

248 When simulating a real oil slick, MEDSLIK-II allows the processes of spreading, evapora-
249 tion, dispersion, emulsification and coastal adsorption to evolve. When the oil first enters the sea,
250 the slick spreads on the sea surface because of gravitational forces. As it is transported, lighter
251 oil components disappear through evaporation and heavier ones emulsify with the water or are
252 dispersed in the water column. Those processes are modelled by means of bulk formulas that
253 needs the oil volume and density as input, as well as sea surface temperature and wind velocity
254 that are provided by the met-ocean models. In the present version of MEDSLIK-II interactions
255 between oil and waves are not considered. Thus, waves are not considered when modelling the
256 oil dispersion (vertically and horizontally) and waves are not dissipated due to the interaction
257 with the oil. MEDSLIK-II is also able to take into account adsorption of oil by the coast should
258 the slick reach it. The full description of the transformation processes formulation can be found
259 in De Dominicis et al. (2013b). Those processes have been switched off when calculating the
260 drifters' trajectories.

261 3.2. Oil spill simulations.

262 Starting from the oil slick information acquired by the CSN-2 (see Tab. 1), met-ocean data
263 available through the MEDESS4MS system have been used as input to MEDSLIK-II to predict
264 the position of the oil slick in the next few hours. The initial shapes of the slick are 7 polygons
265 built around the centre coordinates of each oil slick (see Tab. 1). The volumes of each oil slick
266 have been evaluated starting from the areas provided by the CNS-2 system (see Tab. 1) and
267 assuming for each oil patch an average thickness of 35 microns and oil density of 840 kg/m³.
268 These parameters were chosen, because the oil slick was not observed by satellite the day after,
269 and thus it is realistic to assume that the oil evaporated due to low density (light oil) or/and that
270 the oil slick thickness was too thin to be visible from the satellite. The total amount of oil is then
271 63 m³.

272 The meteorological, ocean and wave models available for the SG1 exercise area are listed in
273 Tab. 4, with their spatial coverage and resolution. All the ocean models provided hourly currents
274 fields. Full description of these models is given in Zodiatis et al. (2016). Among this large
275 dataset of models, we decided to force MEDSLIK-II using all the ocean/wave models available
276 in the area and the following two criteria: (1) oil spill simulations are performed using the highest
277 resolution ocean, wave and wind models available (SIM 1 in Tab. 4); (2) oil spill simulations are
278 performed using ocean and wave models that have been forced by the same wind data and thus,
279 representing coherent systems of modelling products (SIM 2 - 5 in Tab. 4). In this first set of
280 simulations currents and waves have been used to advect the oil slick. The Stokes drift has been
281 calculated from wave statistics provided by the wave model, following Eq. 4. The wind has been
282 used only for the calculation of the weathering processes.

SIM.	Ocean
1	PREVIMER MENOR North Western Mediterranean 1.2 km
2	PREVIMER MENOR North Western Mediterranean 1.2 km
3	WMED Western Mediterranean 3.5 km
4	MFS Mediterranean 6.5 km
5	POSEIDON Mediterranean 10 km
SIM.	Wave
1	CYCOFOS WAM4 Mediterranean 5 km
2	PREVIMER MENOR WW3 Western Mediterranean 10 km
3	MFS WW3 Mediterranean 6.5 km
4	MFS WW3 Mediterranean 6.5 km
5	POSEIDON WAM Cycle 4 Mediterranean 10 km
SIM.	Wind
1	SKIRON Mediterranean 5 km
2	ARPEGE Mediterranean 10 km
3	ECMWF Mediterranean 25 km
4	ECMWF Mediterranean 25 km
5	POSEIDON Mediterranean 5 km

Table 4: Ocean, wave and wind data used in the MEDSLIK-II oil spill simulations

283 3.3. Sensitivity to ocean, waves and wind in drifters simulations

284 Drifters deployed into the oil slick on the 17 May 2014 (see Sect. 2.2), have been used to
 285 examine how they respond to ocean currents, waves and wind. Four set of simulations have been
 286 designed in order to quantitatively evaluate (1) the quality of the currents used, (2) the magnitude
 287 of the wave-induced transport, (3) the sensitivity to different Stokes drift parameterizations and
 288 (4) the wind drag effect in modeling trajectories of drifters with different immersion depths.
 289 Starting from the drifters' deployment positions (Tab. 3), their trajectories have been predicted
 290 for 24 hours in the case of CODE and iSPHERE drifters (since the drifters stayed at sea just for
 291 1 day), and for 48 hours in the case of iSLDMB and MAR/GE-T. This allows us to compare
 292 the behavior of completely immersed drifters (CODE, iSLDMB) and partially immersed drifters
 293 (iSPHERE, MAR/GE-T).

294 The first set of simulations, listed in Tab. 5, focuses on the evaluation of the surface ocean
 295 currents that can be provided by ocean models with different horizontal resolution. In this set of
 296 simulations wave and wind advection are not considered.

297 The second set of simulations, listed in Tab. 6, focuses on the evaluation of the magnitude
 298 of the wave-induced transport. The Stokes drift has been calculated from the significant wave
 299 height, wave mean period, and wave mean direction (see Eq. 4), provided by wave forecasting
 300 systems with different spatial resolution. In order to isolate the effect of the Stokes drift from the
 301 ocean currents advection, all the simulations have been forced with the same current field (the
 302 highest resolution currents available).

303 The possibility to calculate the Stokes drift from wave model data has been recently in-
 304 troduced in MEDSLIK-II and it is worthwhile to compare it with the previous method used, the
 305 Stokes drift from JONSWAP wave spectrum parameterization (De Dominicis et al., 2013a, 2014)
 306 that strongly depends on the wind velocity, as shown in Eq. 3. In all the experiments, the highest
 307 resolution currents available have been used, while the wind forcing has been taken from models

308 with different spatial resolution, see Tab. 7.

309 The last set of simulations have been performed in order to test how the drifters with different
 310 overwater structure respond to the wind. In this case in order to evaluate the effect of the wind
 311 and to be able to compare the results with the simulations performed using the Stokes drift from
 312 wave model data, the same ocean and wave forcing have been used, with the addition of the wind
 313 drag velocity calculated from the highest resolution wind model available, see Tab. 8. The wind
 314 drag velocity, U_D has been calculated using the same leeway parameter γ for all kind of drifters:
 315 we tested $\gamma = 0.01$ which is the upper limit of the range suggested by Röhrs et al. (2012) and
 316 $\gamma = 0.03$ to test the practical 'rule of thumb' widely used in previous studies (see Sect. 3).

SIM.	U_C
1C	PREVIMER NW 1.2 km
2C	WMED 3.5 km
3C	MFS 6.5 km
4C	POSEIDON MED 10 km

Table 5: List of simulations forced by different surface currents.

SIM.	U_C	U_S
1S	PREVIMER NW 1.2 km	CYCOFOS MED 5 km
2S	PREVIMER NW 1.2 km	MFS-WW3 6.5 km
3S	PREVIMER NW 1.2 km	PREVIMER MED 10 km
4S	PREVIMER NW 1.2 km	POSEIDON MED 10 km

Table 6: List of simulations forced by surface currents and Stokes drift from wave model output.

SIM.	U_C	U_J
1J	PREVIMER NW 1.2 km	SKIRON MED 5 km
2J	PREVIMER NW 1.2 km	POSEIDON MED 5 km
3J	PREVIMER NW 1.2 km	ARPEGE MED 10 km
4J	PREVIMER NW 1.2 km	ECMWF 25 km

Table 7: List of simulations forced by surface currents and Stokes drift using JONSWAP formulation based on wind model output.

SIM.	U_C	U_S	U_D
1D/1D3	PREVIMER NW 1.2 km	CYCOFOS MED 5 km	SKIRON MED 5 km
2D/2D3	PREVIMER NW 1.2 km	MFS-WW3 6.5 km	SKIRON MED 5 km
3D/3D3	PREVIMER NW 1.2 km	PREVIMER MED 10 km	SKIRON MED 5 km
4D/4D3	PREVIMER NW 1.2 km	POSEIDON MED 10 km	SKIRON MED 5 km

Table 8: List of simulations forced by surface currents, Stokes drift from wave model output and wind drag velocity: D is for 1% leeway factor and D3 is for 3% leeway factor.

317 Two metrics have been used to quantitatively evaluate the accuracy of the drifter trajectory
 318 simulations. The first metric is the absolute separation distance $d_i(\vec{x}_s(t_i), \vec{x}_o(t_i))$ between the ob-
 319 served and the simulated trajectories, where d_i is the distance at the time t_i between the simulated
 320 drifter position, \vec{x}_s , and the observed positions, \vec{x}_o . The second metric is the Liu and Weisberg
 321 (2011) skill score. It is defined as an average of the separation distances weighted by the lengths
 322 of the observed trajectories:

$$s(t_i) = 1 - \frac{1}{n} \frac{\sum_{t=t_0}^{t_i} d_i(\vec{x}_s(t), \vec{x}_o(t))}{\sum_{t=t_0}^{t_i} l_{oi}(\vec{x}_o(t_0), \vec{x}_o(t))} \quad (9)$$

323 where l_{oi} is the length of the observed trajectory at the corresponding time, t_i , after the de-
 324 ployment time t_0 . Such a weighted average tends to reduce the evaluation errors that may arise
 325 using only the absolute separation distance and n is a tolerance threshold. In this work, as sug-
 326 gested by Liu and Weisberg (2011), we used $n = 1$, this corresponds to a criterion that cumulative
 327 separation distance should not be larger than the associated cumulative length of the drifter tra-
 328 jectory. The higher the s value, the better the performance, with $s = 1$ implying a perfect fit
 329 between observation and simulation and $s \leq 0$ indicating the model simulations have no skill.

330 4. Results

331 4.1. Multi-model simulation of oil slick from satellite

332 All the ocean/wave models available from the MEDESS4MS system in the area have been
 333 used as input to MEDSLIK-II in order to provide the prediction of the transport of the oil slick
 334 observed by the CSN-2 system. The predictions after 3 hrs and 6 hrs have been compared against
 335 the oil samples positions (see Tab. 2) and the drifter deployment positions (see Tab. 3), respec-
 336 tively. Fig. 4 shows the MEDSLIK-II predictions forced with the highest resolution forcing
 337 fields available in the area of the exercise (SIM 1 in Tab. 4), while Fig. 5 and Fig. 6 show
 338 MEDSLIK-II simulations performed using ocean and wave models that have been forced by the
 339 same wind data (SIM 2 - 5 in Tab. 4). Only currents and waves have been used to advect the oil
 340 slick and the Stokes drift has been calculated from wave statistics provided by the wave model.
 341 Wind is used only for calculation of weathering processes

342 It is possible to observe that the PREVIMER 1.2 km (SIM1-2) currents (Fig. 4a-b and Fig.
 343 5a-b) and WMED 3.5 km (SIM3) currents (Fig. 5c-d) are directed north and north-west in the
 344 oil slick area, while the MFS 6.5 km currents (SIM4) are north and north-east and much smaller
 345 with respect to the other models (Fig. 6a-b). The POSEIDON 10 km currents (SIM5) are in the
 346 opposite direction with respect to the rest of the models (Fig. 6c-d).

347 The Stokes drift predicted by CYCOFOS WAM4 (SIM1) is directed south in the oil slick area
 348 (Fig. 4a-b). Instead, all the other wave models predict the Stokes drift to the north and north-east
 349 direction, but the MFS WW3 Stokes drift is smaller (Fig. 6a-b) compared to the other models.
 350 The wind has not been used to advect the oil slick, but only for the transformation processes.
 351 Nevertheless, it is interesting to observe that the direction of the local wind does not always
 352 correspond with the direction of the Stokes drift. Indeed, the waves are not always generated by
 353 local wind conditions and estimating the wave-induced transport just from the wind (for example
 354 with JONSWAP), might lead to a wrong estimation of the direction of the Stokes drift.

355 Although the simulation temporal horizon is short, from Fig. 6 it can be observed that the
 356 predictions done with the low resolution (10 km) POSEIDON currents and waves (SIM 5) do not
 357 overlay the oil samples and drifter deployment positions (Fig. 6c-d), while the predicted oil slicks

358 by MFS currents and MFS WW3 waves (6.5 km) (SIM 4) overlay the oil slick predicted positions
359 (Fig. 6a-b), but the latter are not in the higher concentration core of the oil slick. The same is
360 observed in Fig. 4a-b for the simulations performed with the highest resolution forcings available
361 (SIM1: PREVIMER 1.2 km currents and CYCOFOS WAM4 5 km Stokes drift). Instead, for the
362 simulations SIM2 and SIM3 the drifter deployments positions are in the higher concentration
363 core of the oil slick (Fig. 5b-d). SIM1 and SIM2 share the same current forcing (PREVIMER
364 1.2 km), thus the worse performance observed in SIM1 is most probably due to the Stokes drift
365 provided by CYCOFOS WAM4 model. This is of course valid just for this specific case, but it
366 shows that higher resolution not always means higher accuracy.

367 A more extensive validation of the model performances is done in the next Section through a
368 comparison with drifter trajectories.

369 4.2. *Impact of ocean currents in modelling drifter trajectories*

370 Fig. 7-a shows the 2 CODE drifter trajectories overlaid with the simulated trajectories forced
371 with ocean currents with different resolution. It is shown that the lowest resolution currents
372 (POSEIDON MED 10 km - SIM4C) wrongly reproduce the direction of the currents during the
373 first hours of the simulation, although the correct direction is then recovered, the simulated drifter
374 moves slower than the real one. A general underestimation of the strength of the current is also
375 observed in SIM3C, which is forced by the MFS 6.5 km currents. The better results are observed
376 with the highest resolution currents available in SIM1C (PREVIMER NW 1.2 km) and SIM2C
377 (WMED 3.5 km); in both simulations the direction is correct, and in SIM1C the length of the
378 trajectory is comparable with the real one. The above qualitative considerations are confirmed by
379 the absolute separation distance and skill score calculation. Indeed, Fig. 8 shows that separation
380 distance reaches the highest values for the entire simulation period in the case of SIM4C, and the
381 skill score is lower than 0 (no skill) for up to 14 hours of simulation, when the correct direction
382 of the real drifters is finally recovered. SIM3C shows a higher separation distance (and lower
383 skill) than SIM1C and SIM2C, for the entire simulation period. The performance of SIM1C and
384 SIM2C are comparable, although the PREVIMER NW 1.2 km (SIM1C) seems to perform better
385 toward the end of the simulation period, showing a lower separation distance and higher skill
386 score.

387 The simulated trajectories of the 2 iSPHERE drifters show the same behavior of the CODE
388 simulated drifter trajectories (indeed the release point of the drifters was few meters apart), as
389 shown in Fig. 7-b. As observed before, higher resolution currents (SIM1C and SIM2C) perform
390 better than low resolution currents (SIM3C and SIM4C). However, the quantitative comparison
391 with the observations shows a higher separation distance and lower skill score than in the case
392 of CODE drifters, see Fig. 8. The separation distance is growing during the entire simulation
393 period, up to 18 km in case of SIM4C and SIM3C, while in the case of the CODE drifters it
394 reaches a maximum of 7-8 km.

395 The same considerations can be used in analyzing the iSLDMB and MAR-GE/T simulated
396 trajectories (Fig. 7-c and Fig. 7-d) and the corresponding separation distance and skill score
397 (Fig. 9). The iSLDMB, as expected, shows the same behavior of the CODE drifters, as both are
398 completely submerged, while the MAR-GE/T drifter that is partially overwater behaves similarly
399 to the iSPHERE drifters. The skill score reaches 0.7 in the case of the iSLDMB simulations
400 forced with highest resolution currents (PREVIMER NW 1.2 km), while in the case of MAR/GE-
401 T the skill is 0.5. As it was expected in the case of the iSPHERE and MAR-GE/T, the currents are
402 not sufficient to correctly reproduce the transport of this type of drifters, that may be influenced
403 by the wind on the overwater structure, as better explained in Sect. 4.4.

404 4.3. Impact of waves in modelling drifter trajectories

405 Fig. 10 focuses on the effect of waves on the drifter transport. In order to isolate the ef-
406 fect of waves, all the simulations have been performed using the same current field: the highest
407 resolution currents (PREVIMER NW 1.2 km), which were shown to perform better in the previ-
408 ous set of simulations. All the simulations indicated by S (and shown as continuous lines in Fig.
409 10-11-12) have been performed by adding to the currents the Stokes drift calculated from the
410 wave statistics produced by the wave models with different resolution. Simulations tagged by J
411 (and shown as dashed lines in Fig. 10-11-12) have been performed by adding the Stokes drift
412 calculated by using the JONSWAP spectrum formulation, depending on the wind provided by
413 models with different resolution. By comparing Fig. 10 with Fig. 7, it is evident that by adding
414 the Stokes drift the distance travelled by the simulated drifters is enhanced, this is true for all the
415 drifters, but it is more evident in the 48 hours simulations. As it can be seen in Fig. 10-a and
416 Fig. 10-b and more evidently in Fig. 11, for the first 15 hours the addition of the waves does not
417 produce any effect, since the sea was calm and low wind conditions were experienced. When
418 the wind starts to increase (after 15 hours, see Fig. 13), the effect of the waves become visible
419 on the simulated drifter paths. From the last few hours of the CODE and iSPHERE trajectories
420 (see Fig. 11), it seems that the addition of Stokes drift-JONSWAP perform better than the Stokes
421 drift from wave model statistics, and this is true for both type of drifters. However, much more
422 information can be extracted by the 48 hours simulated trajectories.

423 Fig. 12 shows that for up to 30 hours the simulations with JONSWAP performs better than the
424 one with the Stokes drift from the wave models, but it is not possible to find any clear connection
425 with the resolution of the wind models. Indeed, it seems that higher resolution winds SIM1J
426 and SIM2J (SKIRON MED 5km and POSEIDON MED 5 km) show lower separation distance
427 (higher skill score) than the lower resolution winds SIM3J and SIM4J (ARPEGE 10 km and
428 ECWMF 25 km). In the case of the Stokes drift from the wave models, up to 30 hours, the
429 performance is not connected with the model resolution, but for all simulations the separation
430 distance and skill score are better than with the transport driven just by the ocean currents (the
431 grey line in Fig. 10-11-12). The above considerations are valid for both iSLDMB and MAR-
432 GE/T drifters.

433 In Fig. 12 it is interesting to observe what happens when the wind continues to increase. Af-
434 ter 30 hours from the drifters' deployment high wind speeds are experienced (see Fig. 13) and in
435 the case of the iSLDMB drifters the separation distance is higher by using JONSWAP than with
436 Stokes drift from wave model data. The separation distance decreases with higher resolution
437 wind models (lower separation distance with SIM1J - SKIRON MED 5 km and SIM2J POSEI-
438 DON MED 5 km), as confirmed also by the skill score. Apart from SIM3S (PREVIMER MED
439 10 km), all the wave models show a comparable performance, as it showed by the skill score
440 trend. In the case of MAR-GE/T the separation distance is always lower by using the JONSWAP
441 spectrum parameterization. By looking to the skill score, no correlation is found between the
442 wind resolution and the simulations' accuracy. In the case of the Stokes drift parameterizations,
443 it seems that SIM3S perform better than the other wave models.

444 From the above comparison what is evident is that in high wind speed conditions, the addition
445 of the Stokes drift calculated by the JONSWAP parameterization leads to an overestimation of
446 the displacement of the iSLDMB, thus leading to worse performance. The addition of the Stokes
447 drift from wave model statistics performed better, but is difficult to conclude which is the model
448 with the highest accuracy. In low wind conditions, the addition of the Stokes drift calculated by
449 JONSWAP leads to better results, compensating for the underestimation of the ocean currents.

450 When this compensation is too high due to high wind speeds, this simple additive correction
451 does not work. In the case of the MAR-GE/T the addition of the Stokes drift calculated by
452 JONSWAP always leads to better results, in this case the overestimation of the Stokes drift does
453 not compensate only for the underestimation of the ocean currents, but also might fill the gap of
454 the missing wind drag process, as it is shown in the next Section.

455 4.4. *Impact of wind drag in modelling drifter trajectories*

456 Fig. 14 shows the effect of the wind drag on the different types of drifters. In order to eval-
457 uate the effect of the wind and to be able to compare the results with the simulations performed
458 using the Stokes drift from wave model data, the same ocean and wave forcing have been used,
459 with the addition of the wind drag velocity calculated from the highest resolution wind model
460 available. As discussed in the previous Section, the JONSWAP parameterization may lead in
461 some cases to better results than the Stokes drift from wave model output. However, the JON-
462 SWAP parameterization simply relies on the wind velocity by assuming that wind and waves are
463 aligned and that the waves are generated only by the local wind, something that is not always the
464 reality (swell is not considered). Thus, we believe it is more correct to use the Stokes drift from
465 wave model output. Since we want to focus on the effect of wind in modelling different kinds of
466 drifters, we chose to force all the simulations with the highest resolution wind model available,
467 although in the simulations using JONSWAP it was not possible to determine which wind model
468 performs better.

469 We tested the addition to the currents and Stokes drift of 1% ($\gamma = 0.01$) and 3% ($\gamma = 0.03$)
470 of the wind velocity in the direction of the wind. As shown in Fig. 14, when the wind drag
471 velocity is added to the currents and waves, the simulated drifters transport direction is deviated
472 in the direction of the wind. During the first 15 hours of simulation, this effect is not evident
473 due to the low wind velocity (see Fig. 13). However, as it can be observed in Fig. 15, during
474 the last hours of the simulation the addition of the wind drag velocity leads to an increase of
475 the separation distance in the case of the CODE drifters, which is more evident when using a
476 3% leeway factor. On the other hand, the iSPHERE drifters present a decrease of the separation
477 distance when adding the wind drag velocity and with 3% leeway factor the separation distance
478 is lower than with 1%. As described in Sect. 3.1, in the particular case of the over-water structure
479 and the submerged part of the object being the same, as in the case of the iSPHERE drifter, the
480 parameter γ can indeed be equal to 0.035.

481 The wind drag effect can be better evaluated on the 48 hours drifter simulations. As can be
482 observed in Fig. 14-d, the addition of the wind deviates the MAR-GE/T simulated drifters in
483 the direction of the real drifter path, which is very evident with a 3% leeway factor, while this
484 deviation is not needed for the iSLDDB drifters. As shown in Fig. 16, the simulation with 3%
485 leeway factor performs worse for the iSLDDB after 15 hours, i.e. when the wind velocity starts
486 to increase. On the other hand, up to 30 hours the simulations with 1% leeway factor perform
487 better for both iSLDDB and MAR-GE/T drifters. However, after 30 hours when the wind starts
488 to further increase (see Fig. 13), in the case of the iSLDDB the addition of the 1% or 3% leeway
489 factor produces a decrease of the skill score (and increase of the separation distance), while in
490 the case of the MAR-GE/T the skill score is higher when considering the wind drag velocity. It
491 should be noted, that in this specific case study, the final change in direction of the MAR-GE/T
492 observed drifters' path (see Fig. 14-d) is not reproduced by any ocean, wave or wind model, thus
493 leading to a general worsening of the performances toward the end of the simulation period.

494 As was observed for the Stokes drift JONSWAP parameterization, in the case of low wind
495 conditions the addition of the wind drag velocity with 1% leeway factor leads to better results

496 with all types of drifters. This might be due to an enhancement of the drifters' displacement, that
497 is underestimated by using only the ocean currents and waves. The effect of the wind-driven sea
498 surface currents and subsurface turbulence might not be adequately resolved by most oceanographic
499 models in the uppermost centimeter of water column, leading to an underestimation of
500 the the actual wind induced surface current. However, in the case of high wind speed and/or 3%
501 leeway factor, it is evident that the addition of the wind drag velocity leads to a deviation of the
502 simulated drifters in the direction of the wind that does not happen in the real trajectories of the
503 fully submerged drifters (iSLDMB).

504 5. Conclusions

505 During the 10 days of MEDESS4MS Serious Game 1 exercise, one of the oil slicks that were
506 observed by satellite was effectively found at sea. Samples of oil were collected and drifters
507 with different water-following characteristics were deployed into the oil slick. Although we did
508 not succeed in the collection of a time series of satellite observations of the same oil slick to be
509 compared with the drifter trajectories, the oil slick in-situ observations and drifter trajectories
510 have been used to evaluate the quality of the ocean, wave and meteorological models forecasts
511 that are accessible from the MEDESS4MS system.

512 MEDSLIK-II predictions of the oil slick evolution, using different combinations of ocean-
513 wave-meteorological models, have been compared with the oil observations at sea 3 hrs and 6
514 hrs after the simulation start. Although the simulation temporal horizon is short, we found that
515 low resolution ocean data perform worse than higher resolution ocean models. The same was
516 found from the analysis of the behavior of different types of drifters at sea: all drifters are better
517 reproduced by using higher resolution ocean models. The final objective of this comparison was
518 not to determine which model is the best among the others, but was to show that we should deal
519 with the uncertainties generated by different model outputs. A multi-model approach can help
520 to quantify uncertainties related to the met-ocean fields. When all the models are in agreement,
521 we might be more confident in the accuracy of the forecasts, on the other hand they might also
522 be all affected by the same error. It is thus difficult to deal with different model outputs without
523 an objective method of analyzing them. Very few examples are available on using met-ocean
524 ensembles in Lagrangian trajectory models (Vandenbulcke et al., 2009; Scott et al., 2012; Wei
525 et al., 2013). Those studies demonstrated that the ensemble can generate important uncertainty
526 information, in addition to predicting the trajectory with higher accuracy than a single ocean
527 model forecast. In the future, a method to weight available ocean/met/wave/oil spill forecasts
528 against validation metrics in order to provide an estimate of the confidence level of each member
529 of the multi-model ensemble has to be developed. The final aim has to be a tool that will be
530 able to compile all the collected results from the different models and produce a synthetic output
531 (such as the probability density charts), that could be used by the end-users.

532 Drifters are the most common instruments used for validation of oil spill and/or trajectory
533 models; this study highlighted that we must carefully consider which kind of drifters we are using
534 to validate trajectory simulations, in order to add the correct terms in the trajectory transport
535 equation. All CODE-type drifters (two CODEs and one iSLDMB) are completely submerged
536 and have the same behavior at sea, shown to be mainly driven by surface ocean currents. While
537 one MAR/GE-T and two iSPHERE, that are partially emerged, move similarly and we found
538 that the surface ocean currents are not sufficient to correctly reproduce their transport. It is worth
539 pointing out that two CODE drifters were moving together, and likewise for the two iSPHERE
540 drifters. This allows us to be more confident in saying that the different behavior is due to the

541 different drifter shapes, rather than to sub-mesoscale ocean dynamics. From the analysis of
542 the behavior of different type of drifters at sea, we found that all drifter trajectories are better
543 reproduced by using higher resolution ocean models. The Stokes drift generally enhances the
544 simulated drifter displacements. We found that the JONSWAP parameterization for Stokes drift
545 calculation leads to an overestimation of the displacement, particularly evident with CODE-type
546 drifters and in high wind speed conditions. This overestimation is not evident in the MAR/GE-T
547 and iSPHERE drifters, since it is probably masked by the missing wind drag effect acting on the
548 overwater drifter structure. We found that in the case of low wind conditions the addition of wind
549 drag velocity with 1% or 3% leeway factor leads to better results with all type of drifters. We
550 think that is not due to a real direct wind drag acting on the drifters, but it is most probably due
551 an incorrect reproduction of the wind-driven sea surface currents and subsurface turbulence at
552 the ocean surface by oceanographic models. This is due to resolution constraints, since surface
553 currents provided by an ocean model are actually the currents in the top meter of the water
554 column and due to missing physics describing the mixed turbulent layer at the air/sea interface.
555 On the other hand the addition of a wind drag velocity with 1% or 3% leeway factor in high wind
556 speed conditions leads to a lower skill in the case of submerged drifters (CODE or iSLDMB),
557 while MAR/GE-T and iSPHERE generally are better reproduced with a higher leeway factor.
558 Indeed, we found that the addition of the wind drag velocity leads to a deviation of the simulated
559 drifters in the direction of the wind that has been found to affect only the partially emerged
560 drifters, while the wind drag effect does not affect the fully submerged drifters. This is more
561 evident in high wind speed conditions.

562 In the future it might be interesting to further explore the wave-induced transport term. First,
563 the effect of having the Stokes drift calculated by integration of the full wave spectrum done
564 internally by the wave model, instead of obtaining it a posteriori from bulk wave parameters,
565 should be examined. As shown by Tamura et al. (2012), this might enhance the magnitude of
566 the Stokes drift. Second, by using fully coupled wave-hydrodynamic models it will be worth
567 to estimate the effect of wave-induced currents (Smith, 2006; McWilliams et al., 2004; Mellor,
568 2003, 2008; Ardhuin et al., 2008) on tracer transport.

569 In the future, further experiments are still needed to assess which drifter behaves most sim-
570 ilarly to an oil slick and under which ocean currents and wind conditions. However, oil slicks
571 do not resemble objects with an overwater structure, that feel that wind drag effect and, thus, we
572 may believe oil slicks would behave more like submerged drifters. On the other hand, an oil slick
573 at the air/sea interface is driven by the currents in the top millimeters of the water column, which
574 are certainly linked with wind and wave-induced turbulence, which are still poorly understood
575 and further fundamental research is needed to achieve a full comprehension of the processes
576 acting at the air/sea interface.

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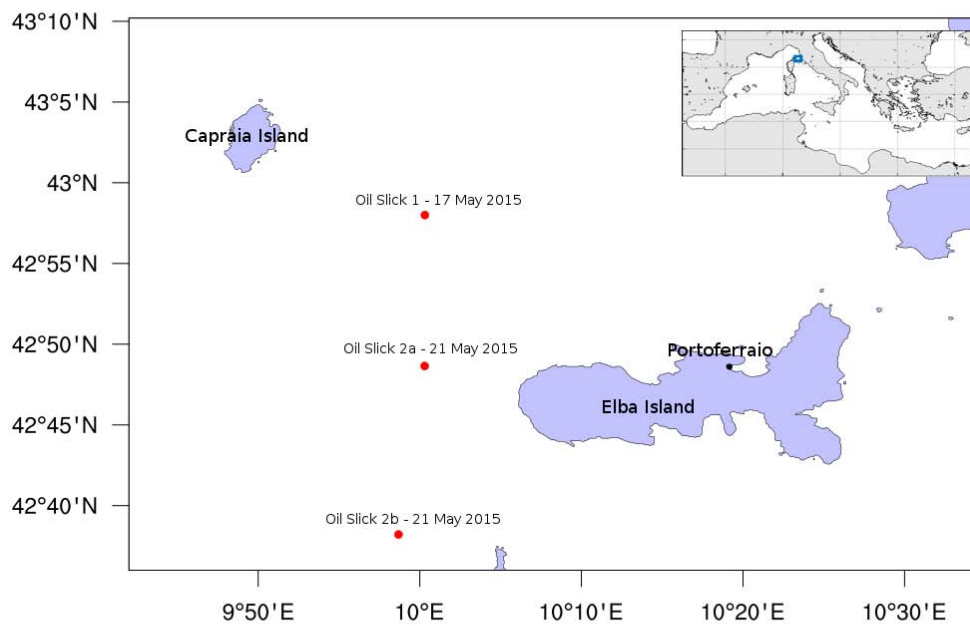
685 **Figures**

Figure 1: MEDESS4MS Serious Game 1 (SG1) exercise area, Elba Island region, Western Mediterranean Sea (red dots are the positions of the oil slicks observed by satellite during the exercise period).

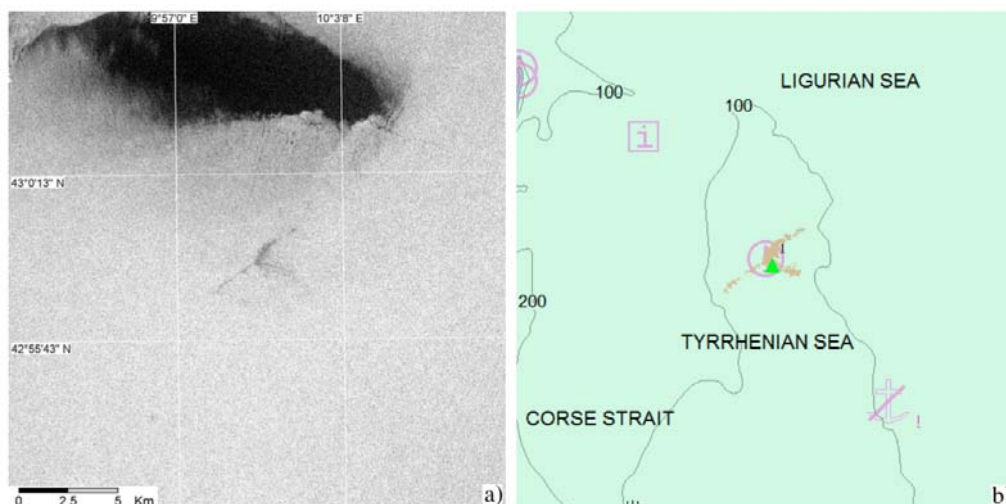


Figure 2: Initial position of the oil slick observed by satellite on the 17th of May 2014 at 05:38 UTC, from EMSA Clean Sea Net alert report received by ITCG. Panel a shows the original satellite image. Panel b is the output of the CSN-2 automatic detection algorithm (the green triangle is the oil slick barycentre).

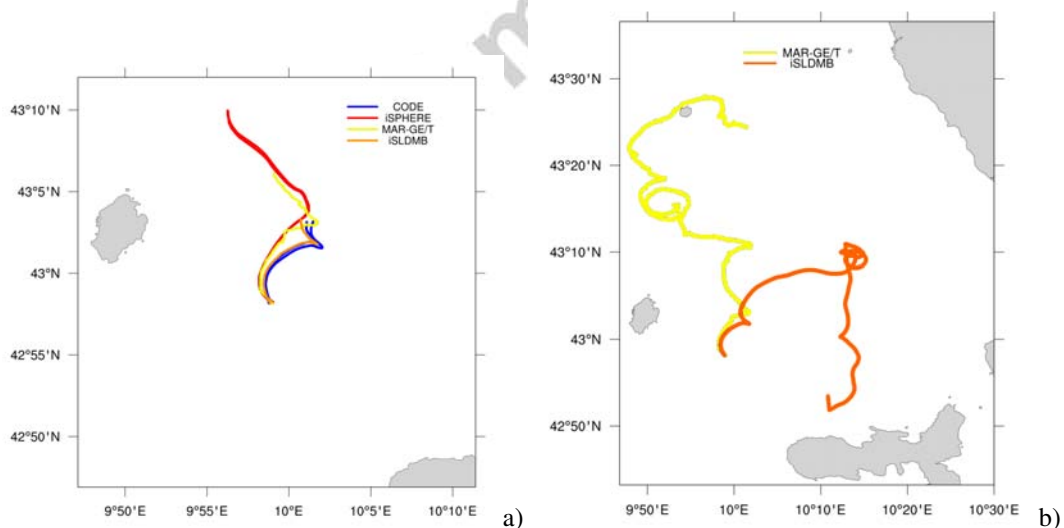


Figure 3: Observed trajectories of the drifters deployed on the 17 May 2014: 2 iSPHERE (red lines) 2 CODE (blue lines), iSLDMB (orange line) and MAR-GE/T (yellow line). Panel a shows the trajectories after 1 day at sea. Panel b shows the trajectories after 7 days at sea (only iSLDMB and MAR-GE/T, since iSPHEREs and CODEs have been recovered after 1 day).

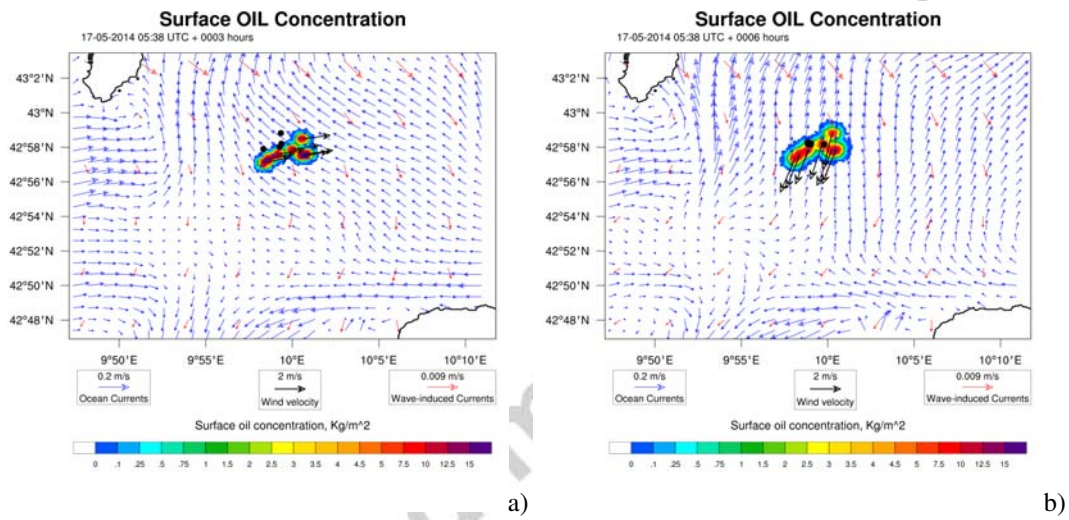


Figure 4: MEDSLIK-II oil slick simulated position performed with the highest resolution forcings available in the area: surface currents from PREVIMER NW 1.2 km, Stokes drift from CYCOFOS WAM4 5 km, winds from SKIRON 5 km. The predicted oil slick positions are compared with in-situ oil sampling positions (black dots in panels a) collected after 3 hours from satellite observations and drifter deployment positions after 6 hours from satellite observations (black dots in panels b). Wind is used only for calculation of weathering processes.

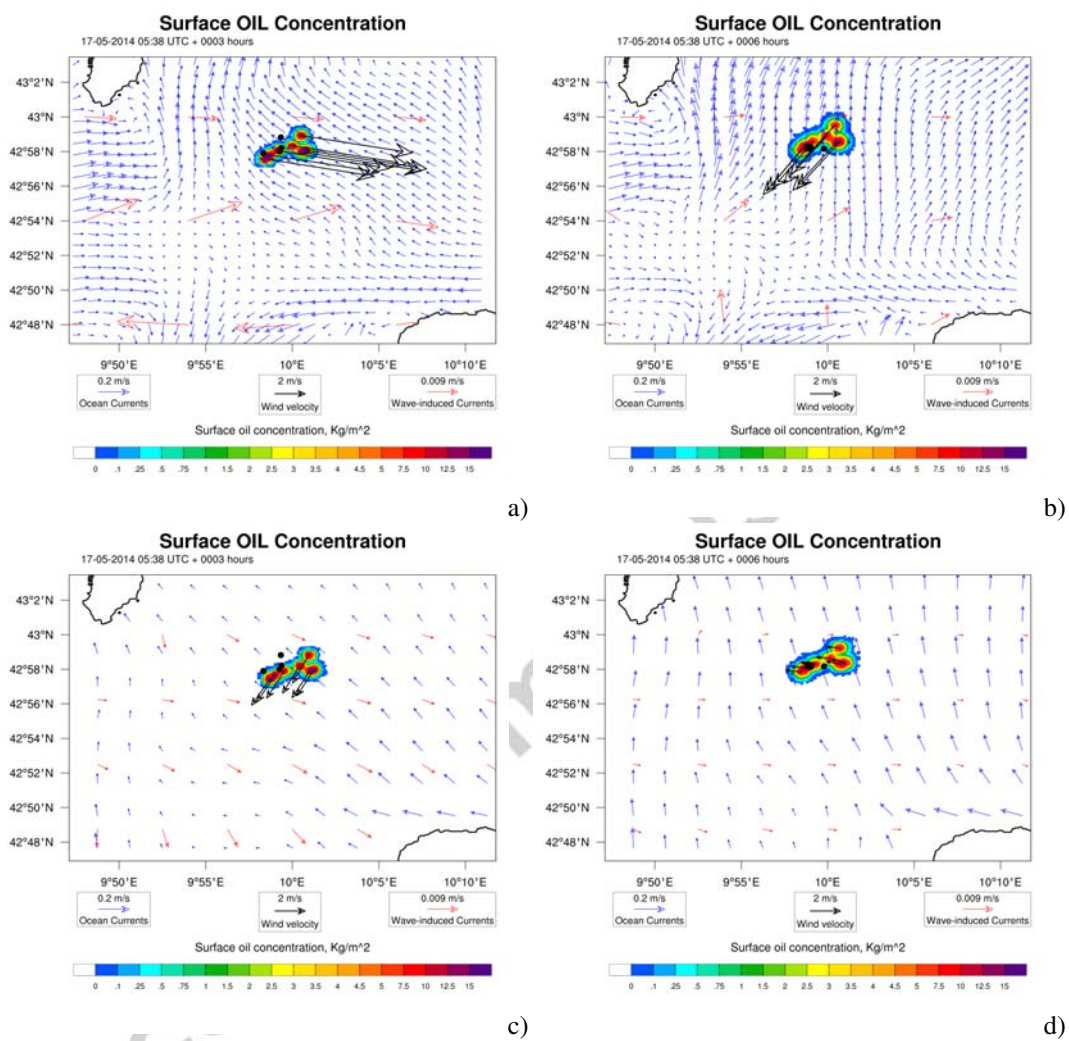


Figure 5: MEDSLIK-II oil slick simulated position compared with in-situ oil sampling positions (black dots in panels a,c) collected after 3 hours from satellite observations and drifter deployment positions after 6 hours from satellite observations (black dots in panels b,d). Panel a-b: simulations have been performed using surface currents from PREVIMER NW 1.2 km, Stokes drift from PREVIMER MED 10 km, winds from ARPEGE MED 10 km; Panel c-d: simulations have been performed using surface currents from WMED 3.5 km, Stokes drift from MFS-WW3 6.5 km, winds from ECMWF 25 km. Wind is used only for calculation of weathering processes.

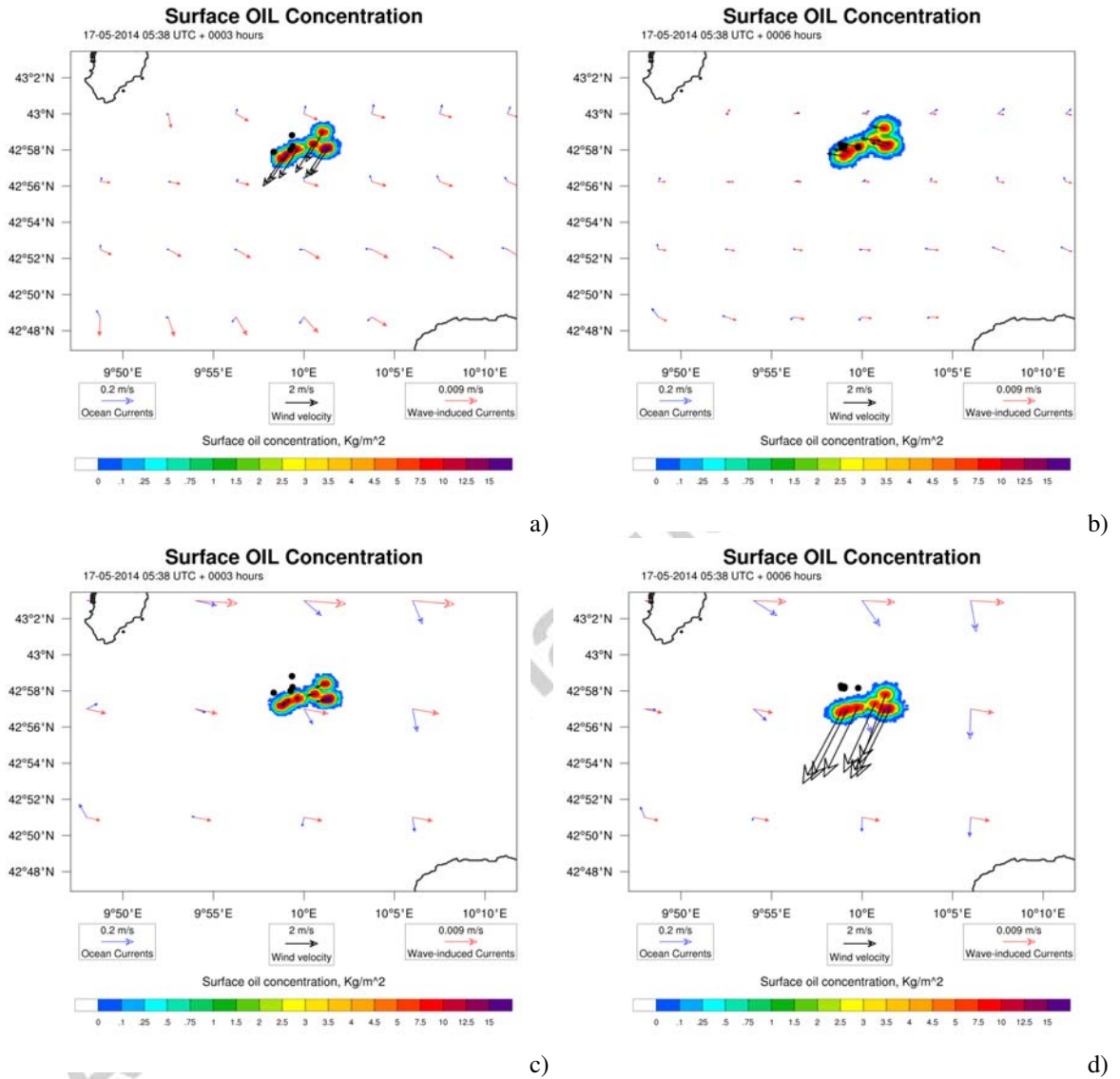


Figure 6: MEDSLIK-II oil slick simulated position compared with in-situ oil sampling positions (black dots in panels a,c) collected after 3 hours from satellite observations and drifter deployment positions after 6 hours from satellite observations (black dots in panels b,d). Panel a-b: simulations have been performed using surface currents from MFS 6.5 km, Stokes drift from MFS-WW3 6.5 km, winds from ECWMF 25 km; Panel c-d: simulations have been performed using surface currents from POSEIDON MED 10 km, Stokes drift from POSEIDON MED 10 km, winds from POSEIDON MED 5 km. Wind is used only for calculation of weathering processes.

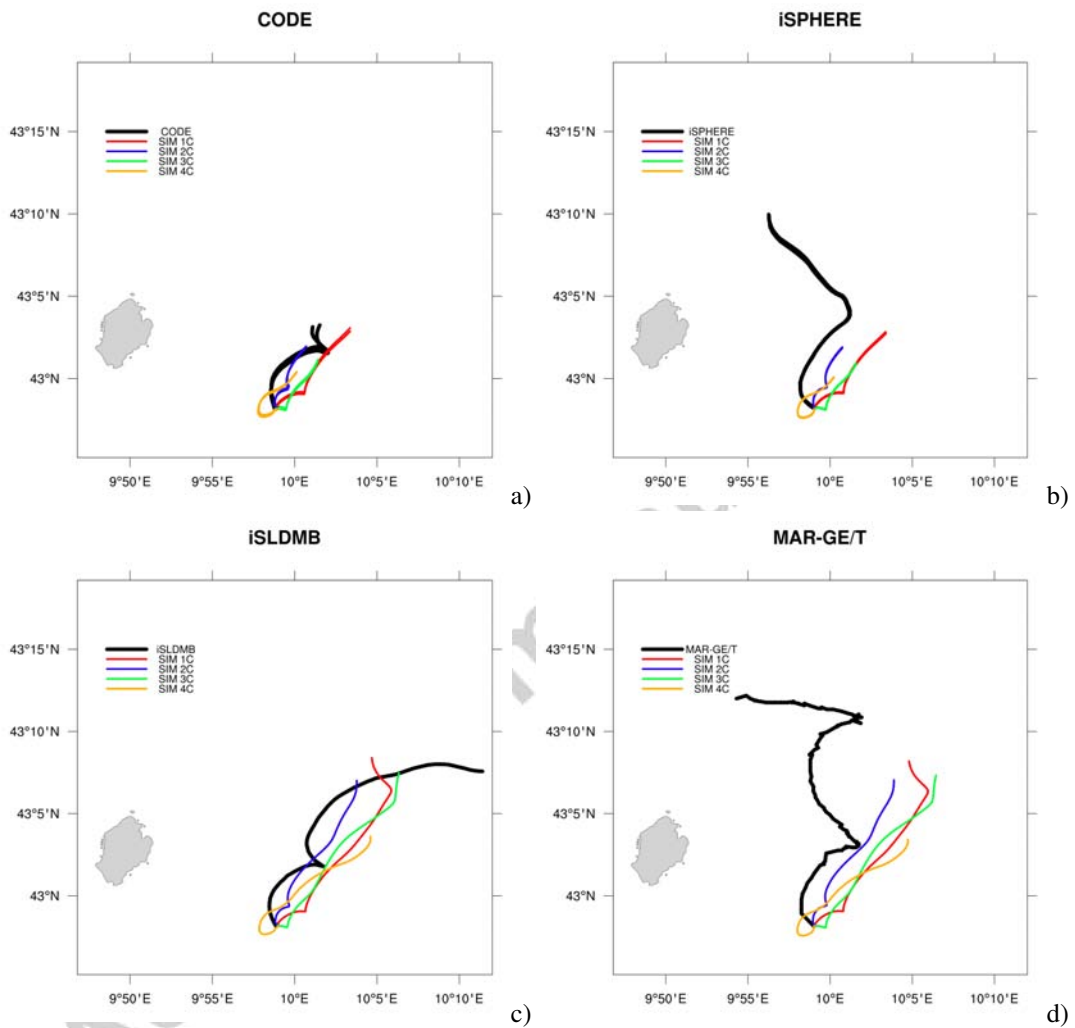


Figure 7: MEDSLIK-II simulated trajectories forced by currents from ocean model listed in Tab. 5: a) CODE 24 hours trajectories; b) iSPHERE 24 hours trajectories; c) iSLDMB 48 hours trajectories; d) MAR-GE/T 48 hours trajectories. The drifter trajectories are the track of the barycentre of the particle cloud displacement, simulated with random walk procedure.

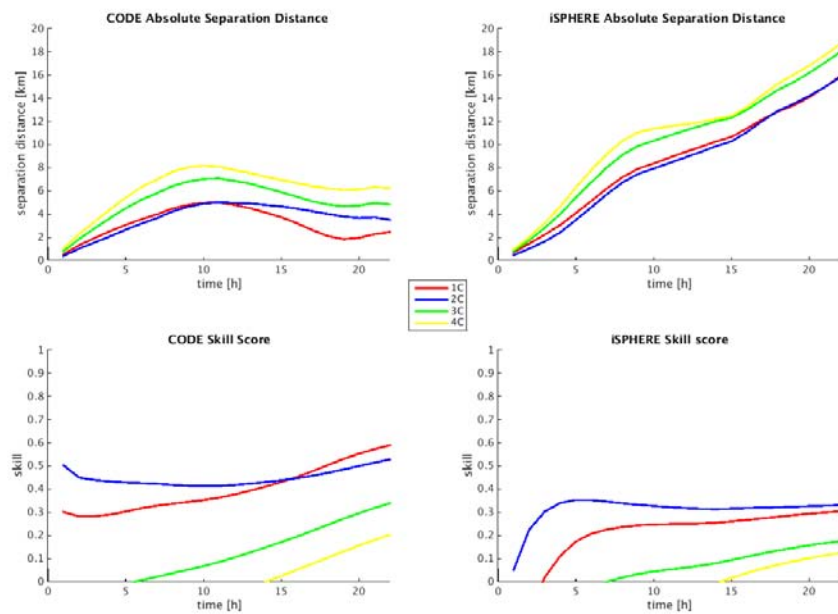


Figure 8: Absolute separation distances and skill scores as a function of the prediction time, for simulations listed in Tab. 5 for CODE and iSPHERE 24 hours trajectories.

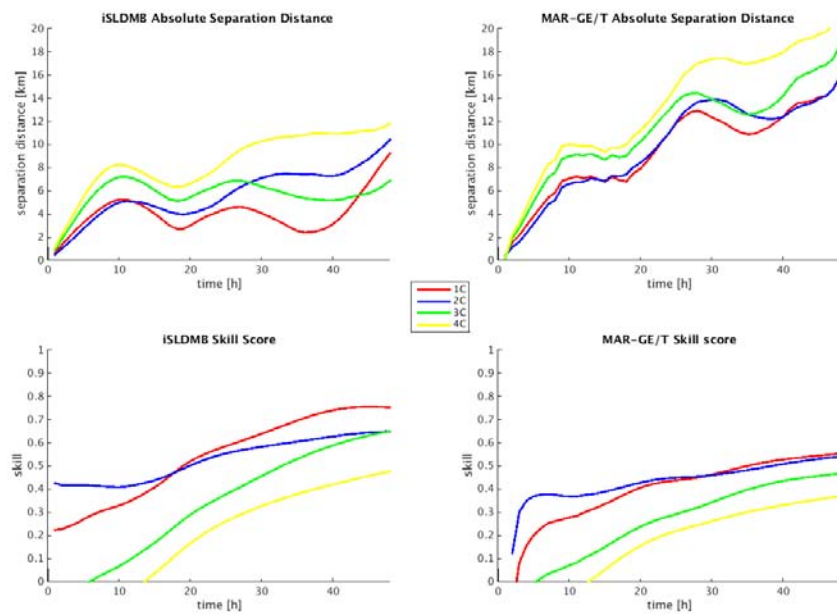


Figure 9: Absolute separation distances and skill scores as a function of the prediction time, for simulations listed in Tab. 5 for iSLDMB and MAR-GE/T 48 hours trajectories.

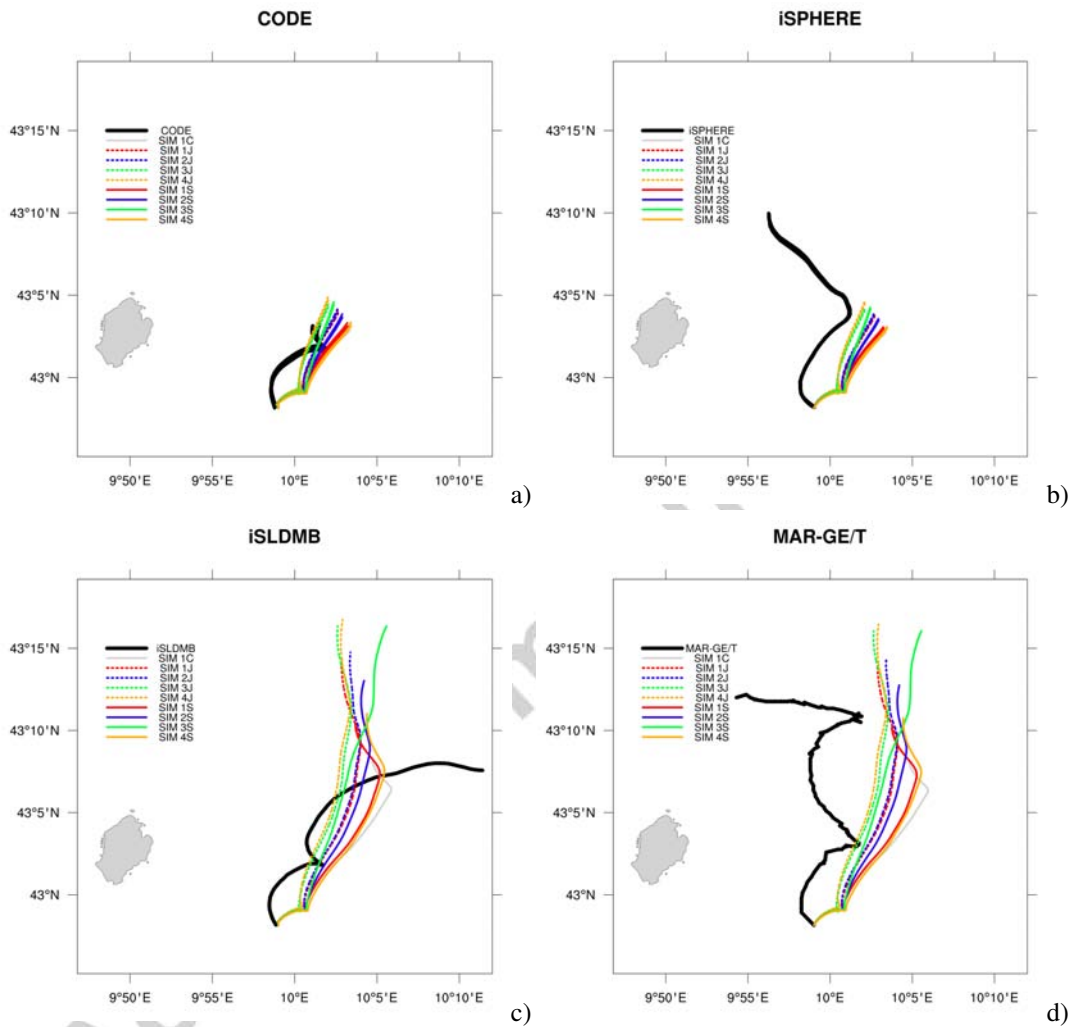


Figure 10: MEDSLIK-II simulated trajectories forced by currents with the addition of the Stokes drift provided by wave models or by JONSWAP parameterization, as listed in Tab. 6 and Tab. 7: a) CODE 24 hours trajectories; b) iSPHERE 24 hours trajectories; c) iSLDMB 48 hours trajectories; d) MAR-GE/T 48 hours trajectories. The drifter trajectories are the track of the barycentre of the particle cloud displacement, simulated with random walk procedure.

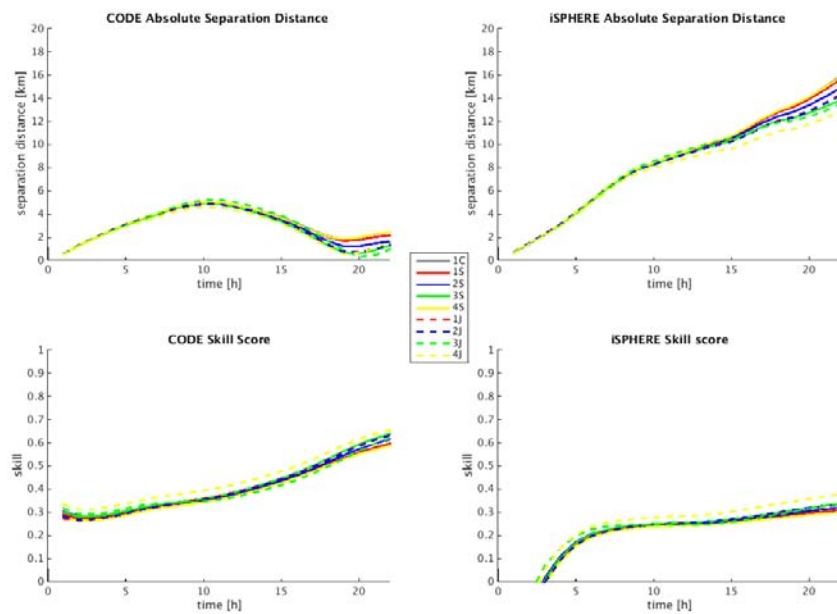


Figure 11: Absolute separation distances and skill scores as a function of the prediction time, for simulations listed in Tab. 6 and in Tab. 7 for CODE and iSPHERE 24 hours trajectories.

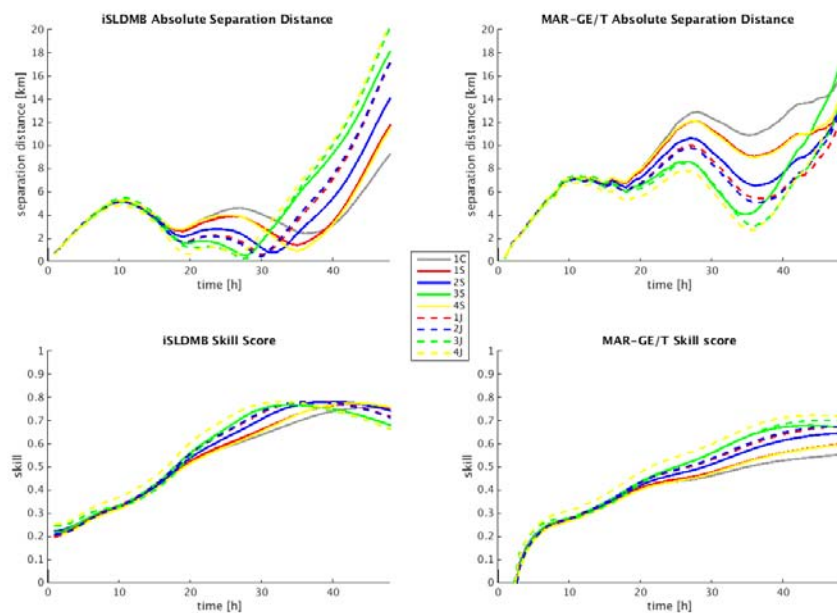


Figure 12: Absolute separation distances and skill scores as a function of the prediction time, for simulations listed in Tab. 5 for iSLDMB and MAR-GE/T 48 hours trajectories.

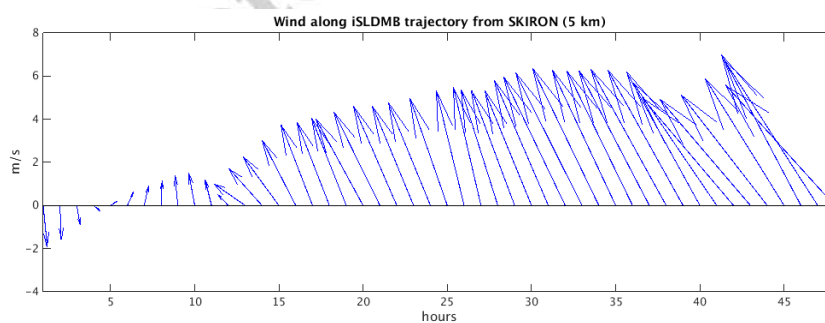


Figure 13: Wind along the iSLDMB drifter trajectory for 48 hours after the deployment (17 May 2014 at 12:00 UTC), from SKIRON wind model (5 km horizontal spatial resolution). The wind along CODE, SPHERE, MAR-GE/T shows the same pattern (not shown).

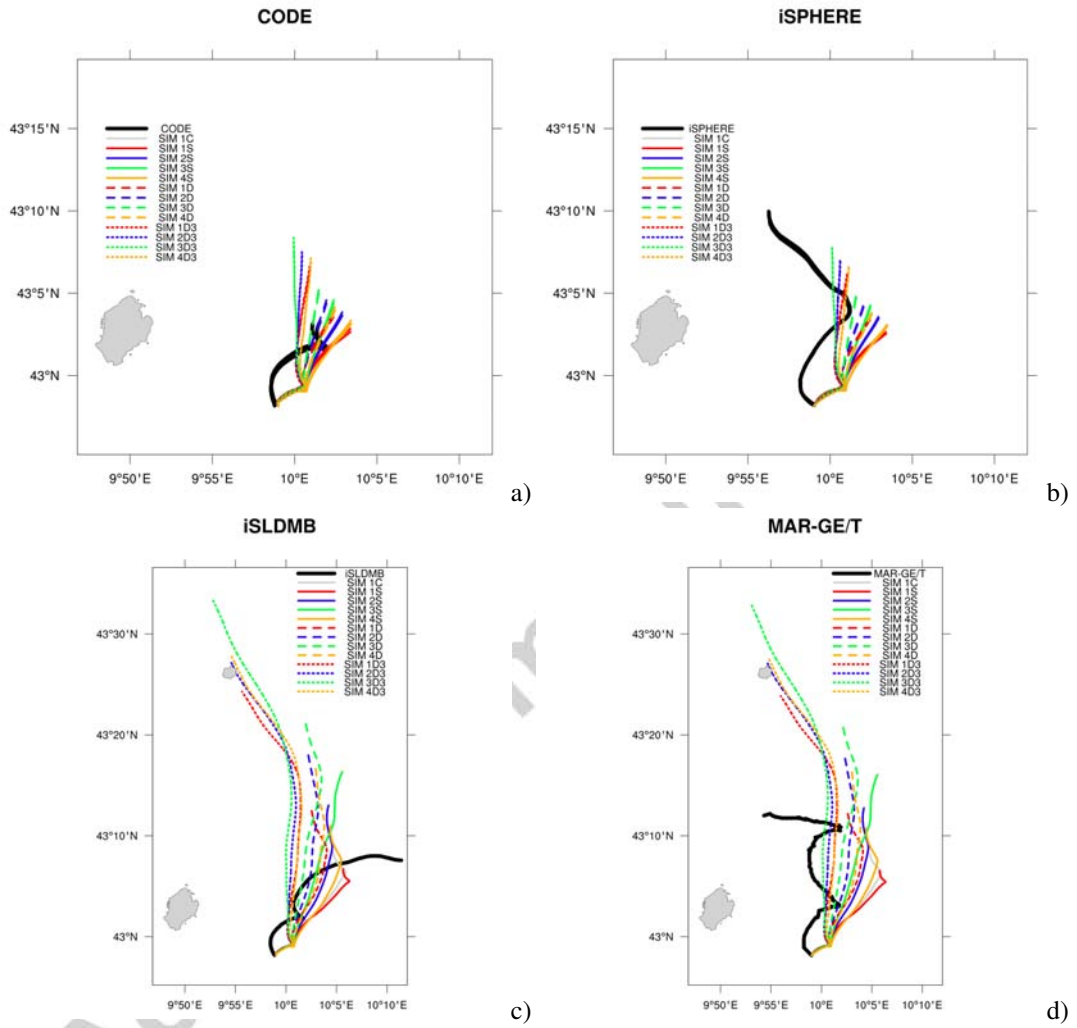


Figure 14: MEDSLIK-II simulated trajectories forced by currents, Stokes drift from wave models and wind drag velocity, as listed in Tab. 8: a) CODE 24 hours trajectories; b) iSPHERE 24 hours trajectories; c) iSLDMB 48 hours trajectories; d) MAR-GE/T 48 hours trajectories. The drifter trajectories are the track of the barycentre of the particle cloud displacement, simulated with random walk procedure.

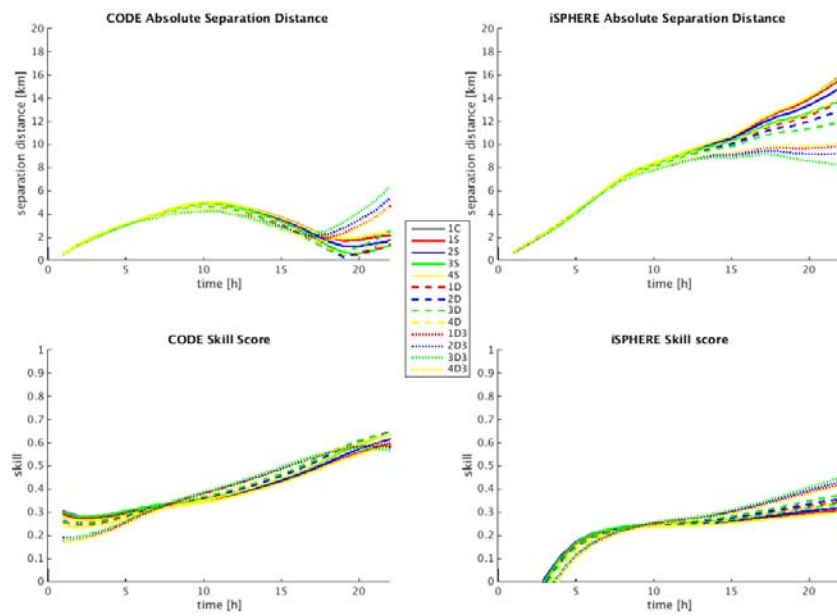


Figure 15: Absolute separation distances and skill scores as a function of the prediction time, for simulations listed in Tab. 8 for CODE and iSPHERE 24 hours trajectories.

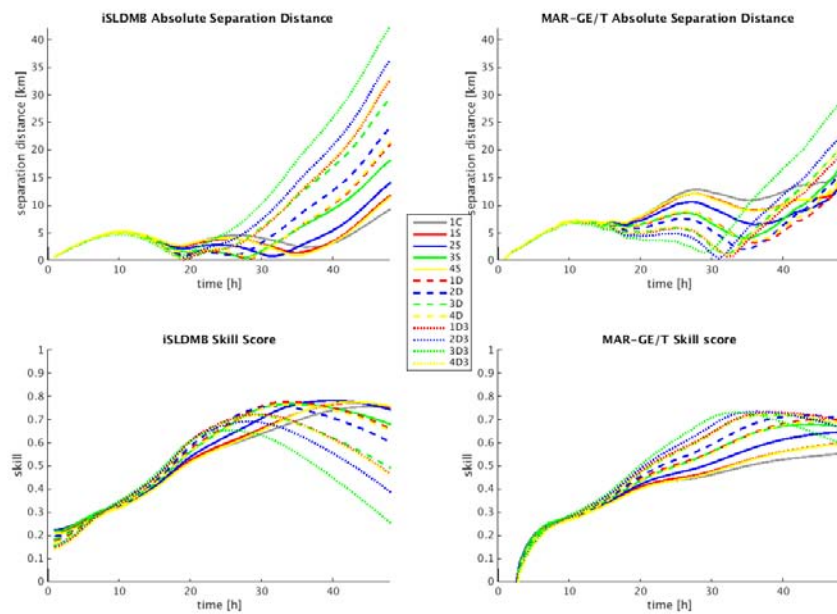


Figure 16: Absolute separation distances and skill scores as a function of the prediction time, for simulations listed in Tab. 8 for iSLDMB and MAR-GE/T 48 hours trajectories.