

Title:

- 1 MERIS-based ocean colour classification with the discrete Forel-
- 2 Ule scale

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9 **Abstract**

10 Multispectral information from satellite borne ocean colour sensors is at present used to
11 characterize natural waters via the retrieval of concentration of the three dominant optical
12 constituents; pigments of phytoplankton, non-algal particles and coloured dissolved organic
13 matter. A limitation of this approach is that accurate retrieval of these constituents requires
14 detailed local knowledge of the specific absorption and scattering properties. In addition, the
15 retrieval algorithms generally use only a limited part of the collected spectral information. In
16 this paper we present an additional new algorithm that has the merit to use the full spectral
17 information in the visible domain to characterize natural waters in a simple and globally valid
18 way. This Forel-Ule MERIS (FUME) algorithm converts the normalized multi-band reflectance
19 information into a discrete set of numbers using uniform colourimetric functions. The Forel-Ule
20 scale is a sea colour comparator scale that has been developed to cover all possible natural sea
21 colours, ranging from indigo blue (the open ocean) to brownish-green (coastal water) and even
22 brown (humic-acid dominated) waters. Data using this scale have been collected since the late
23 nineteenth century, and therefore, this algorithm creates the possibility to compare historic
24 ocean colour data with present-day satellite ocean colour observations. The FUME algorithm
25 was tested by transforming a number of MERIS satellite images into Forel-Ule colour index
26 images and comparing in situ observed *FU* numbers with *FU* numbers modelled from in situ
27 radiometer measurements. Similar patterns and *FU* numbers were observed comparing MERIS
28 ocean colour distribution maps with ground truth Forel-Ule observations. The *FU* numbers

29 modelled from in situ radiometer measurements showed a good correlation with observed *FU*
30 numbers ($R^2 = 0.81$ when full spectra are used and $R^2 = 0.71$ when MERIS bands are used).

31

32

33 **1 Introduction**

34 The application of optical satellite remote sensing techniques to monitor the radiation scattered back
35 from the water column became a major breakthrough in the seventies for monitoring ocean, sea and
36 coastal areas (IOCCG, 1998). Dedicated ocean colour instruments, like CZCS, SeaWiFS, MERIS and
37 MODIS-AQUA, have provided fundamental new insight in the dynamics and role of oceanic plankton
38 (e.g. Behrenfeld et al., 2006). Observations are now starting to span multiple decades, allowing a first
39 glimpse at long-term variations in the plankton composition of the oceans, which are potentially
40 related to global change (Antoine et al., 2005; Polovina et al., 2008).

41 With the launch, in 2002, of the MERIS instrument (Rast et al., 1999) that measures water-leaving
42 reflectance in fifteen spectral bands with high signal-to-noise, it became possible to collect water-
43 leaving radiance with high confidence in regional seas and coastal waters. This has led to development
44 of many new algorithms that can retrieve not only the phytoplankton pigments, but also the mass
45 concentration of suspended material and the absorption by dissolved material (Van der Woerd and
46 Pasterkamp, 2008; Odermat et al., 2012). These algorithms are either simple, calibrated at the local
47 water constituents, or complex with a need for detailed measurements of the specific absorption and
48 scattering properties of these in-water constituents (see e.g. Tilstone et al., 2012). The derived water-

49 quality parameters are the major products of ocean-colour instruments, while the colour itself can be
50 considered as a primary product.

51 Long before the development of diode-arrays to measure spectral radiation, another method had
52 been developed and tested which recorded the colour of natural waters. Towards the end of the
53 nineteenth century, Forel and Ule (Forel, 1890; Ule 1892) proposed a method to classify the colour of
54 the oceans, regional seas and coastal waters using a colour comparator scale. The scale became
55 known as the Forel-Ule (*FU*) scale and since then scale observations have been performed, generating
56 hundreds of thousands of data points over a globe-scale for more than a century. Recently, it was
57 shown (Wernand and Van der Woerd, 2010a) that the *FU* scale is can be used to characterize the
58 colour of natural waters. More importantly, the analysis of *FU* colour variation in the North Pacific
59 since 1930 has revealed significant variations at decadal timescales (Wernand and Van der Woerd,
60 2010b).

61 In this paper we describe a simple algorithm to couple historically collected ocean colour data,
62 obtained over a long time span, with presently available satellite-derived ocean colour imagery
63 for hind casting long-term changes. This Forel-Ule to MERIS (FUME) algorithm converts MERIS
64 observations of sea- and ocean colour to chromaticity coordinates and subsequently to a
65 discrete Forel-Ule number. This will result in a new MERIS water quality product that can be
66 used as a simple and straight-forward index of water colour in addition to the water-quality
67 parameters that are retrieved by inversion schemes. Based on the FUME product, ocean colour
68 trends can be constructed, reaching back to over one hundred years. Distinct optical water
69 types can now be classified according to Forel-Ule's scale and makes it possible to enhance

70 satellite derived products, such as chlorophyll (Moore et al., 2009). Ocean color remote sensing
71 techniques have traditionally been based on two optical water types, known as “Case 1” and
72 “Case 2” (Morel & Prieur, 1977). However, this classification is mainly based on the intrinsic
73 composition, i.e. the role of algae (and related degradation products) in the generation of water
74 colour.

75 Moore et al. (2009) proposed to extend the optical water classification to eight clusters, based
76 on an unsupervised classification of the NOMAD data base of remote sensing reflectance
77 spectra. The reflection spectrum of each satellite pixel has a certain probability to belong to
78 each of the 8 clusters. Another classification method that can be tuned to local properties is
79 proposed by Hommersom et al. (2011). In this work we go back to use the oldest classification
80 of 21 pre-defined scales and use the relative colour difference (colour comparator scale)
81 instead of absolute remote sensing reflectance to classify each pixel to only one representative
82 FU-number.

83

84

85 **2 Methods**

86 In this section we introduce the MERIS satellite data, the algorithm to convert MERIS reflection data to
87 *FU* numbers and the ship-borne measurements for a first characterisation of the FUME results.

88

89 **2.1 MERIS products**

90 MERIS is a 68.5° field-of-view push-broom imaging spectrometer (Rast et al., 1999) on the
91 ENVISAT platform. It measures the solar radiation reflected by the ocean at a spatial resolution
92 of 260 m × 290 m in 15 spectral bands. The bands are programmable in width and position, at
93 visible and near infrared wavelengths. MERIS provides global coverage in 3 days with radiation
94 reflected by the ocean that is atmospherically corrected to derive the normalized water-leaving
95 reflectances, a MERIS Level 2 product (ESA, 2012). The atmospheric correction assumes that the
96 water totally absorbs the NIR, but also includes a correction for those sediment loaded waters
97 where this assumption fails. Normalized water-leaving reflectance (dimensionless) $[\rho_w]_N$ is
98 defined by Eq. 1 in as:

99
$$[\rho_w]_N(\lambda) = \frac{[L_w]_N(\lambda)}{F_0(\lambda)} \tag{1}$$

100 where $[L_w]_N$ is the normalized water-leaving radiance (Gordon and Voss, 1999) and F_0 is the
101 extraterrestrial solar irradiance at wavelength (λ). In this analysis, data is limited to the visible
102 spectrum, covering the first nine MERIS bands with a bandwidth of 10 nm, except for Band 8
103 that has a bandwidth of 7.5 nm (Table 1). In the standard processing by ESA a number of global
104 products are derived together with $[\rho_w]_N$ that will be used to compare *FU* products with
105 standard ESA products: Algal-1 and Algal-2, the index for Chlorophyll-a concentration in case-1
106 and case-2 waters, respectively, SPM (suspended particulate matter) and YS (an index for
107 absorption by dissolved matter). For documentation and additional references we recommend
108 the MERIS algorithm theoretical baseline document (ESA, 2012).

109

110 **2.2 The FUME algorithm**

111 The FUME algorithm converts the normalized water-leaving reflectance in nine MERIS bands
112 into a discrete *FU* number in three steps: Step 1: Calculation of the tristimulus values *X*, *Y*, *Z* by
113 calculating the convolution of the Colour Matching Functions (CMF's) and the normalized
114 water-leaving reflectance (CIE, 1932). Step 2: Calculation of the (*x*, *y*) chromaticity coordinates
115 by the ratio of *X* or *Y* tristimulus value and the sum of the tristimulus values. Step 3:
116 Determination of the *FU* scale number by comparison of calculated (*x*, *y*) values to the unique
117 chromaticity coordinates of the twenty-one *FU* numbers.

118

119 Step 1: Tristimulus values are the amounts of three primaries that specify a colour stimulus of
120 the human eye (Wyszecky and Stiles, 2000) and are noted as *X*, *Y* and *Z* (CIE, 1932). The CIE
121 1931 standard colourimetric 2 degree CMFs \bar{x} (red), \bar{y} (green) and \bar{z} (blue) are presented in
122 Fig. 1. These serve as weighting functions for the determination of the tristimulus values of the
123 MERIS normalized water-leaving reflectance $[\rho_w]_N$ by the Eq. 2a, 2b and 2c;

$$124 \quad X = \int [\rho_w]_N(\lambda) \bar{x}(\lambda) d\lambda \quad (2a)$$

$$125 \quad Y = \int [\rho_w]_N(\lambda) \bar{y}(\lambda) d\lambda \quad (2b)$$

$$126 \quad Z = \int [\rho_w]_N(\lambda) \bar{z}(\lambda) d\lambda \quad (2c)$$

127

128 Because MERIS does not provide full-spectral coverage, the reflection spectrum is first
 129 reconstructed by linear interpolation between band n=1 (412.5 nm) and band n=9 (708 nm)
 130 with a resolution of 1 nm. An example is shown as black line in Fig. 1. Note that the linear
 131 interpolation at λ_i (nm) is always carried out between subsequent bands (n, n+1) with the
 132 condition ($\lambda_n < \lambda_i < \lambda_{n+1}$). The tristimulus values for X, Y and Z are obtained by a Riemann sum
 133 approximation of the integrals with $\Delta\lambda= 1$ nm resolution;

134

$$135 \quad X = \sum_{i=413}^{708} [\rho_w]_N(\lambda_i) \bar{x}(\lambda) \Delta\lambda \quad (3a)$$

136

$$137 \quad Y = \sum_{i=413}^{708} [\rho_w]_N(\lambda_i) \bar{y}(\lambda) \Delta\lambda \quad (3b)$$

138

$$139 \quad Z = \sum_{i=413}^{708} [\rho_w]_N(\lambda_i) \bar{z}(\lambda) \Delta\lambda \quad (3c)$$

140

141

142 Step 2: Subsequently the chromaticity coordinates x, y and z are calculated from the ratio of
 143 each of the tristimulus values and the sum of the values:

$$144 \quad x = \frac{X}{X+Y+Z} \quad y = \frac{Y}{X+Y+Z} \quad z = \frac{Z}{X+Y+Z} \quad (4)$$

145 As $x+y+z = 1$, and therefore $z = 1-x-y$, the third coordinate offers no additional information and
 146 only two coordinates (by convention x and y) are used to represent the colour in a so-called
 147 chromaticity diagram (see e.g. Mobley, 1994). The white point W in the chromaticity diagram

148 has the coordinates $x=y=z=1/3$ (Fig. 2). The ratio of the distance between W and an arbitrary
149 point P (a) and the distance from W to the spectral locus ($a+b$), gives the colour saturation
150 ($a/(a+b)$) or the intensity of the colour in P. In this way, the chromaticity coordinates (x_M, y_M)
151 for every MERIS pixel can be calculated.

152

153 Step 3: In the next step the (x_M, y_M) is converted to a *FU* number. The original *FU* scale was
154 created to make an objective classification of natural waters (see for a review Wernand and
155 Gieskes, 2011). In 21 glass tubes a variable mixture of three standard solutions (distilled water,
156 ammonia, copper-sulphate, potassium-chromate and cobalt-sulphate) were created to obtain
157 the colour-palettes of the scale. These standard solutions were recently reconstructed and their
158 optical properties were measured in the laboratory with medium resolution spectrometers
159 (Wernand and Van der Woerd, 2010a). The calculated chromaticity coordinates of the original
160 *FU* scale are presented in Table 2 and graphically shown as a line of black dots, between the
161 white point and the locus, in the chromaticity diagram of Fig. 2. The FUME algorithm first shifts
162 the origin to the white point W with chromaticity coordinates $x_W = y_W = 1/3$ (Fig. 3). Then it
163 calculates the angle (α_M) between the vector to a point with certain *FU* coordinates (x_M, y_M) and
164 the positive x-axis (at $y-y_w = 0$), giving higher angles in an anti-clockwise direction and compares
165 these with the angles (α_i) of the *FU* solutions (Table 3). All calculations were made with the
166 ATAN2 function (four-quadrant inverse tangent) and the derived angles (in radials) were
167 multiplied by $180/\pi$ to get the angles in degrees:

168
$$\alpha_M = \arctan(y_M - y_W, x_M - x_W) \text{ modulus } 2\pi \quad (5)$$

169

170 Two examples are shown in Fig. 3: α_i is the angle matching *FU* number 8. The yellow dot is
171 derived from the normalized spectral reflectance of a MERIS-pixel and coordinates ($x_M - x_W = -$
172 0.15, $y_M - y_W = 0.1$) and angle (α_M). Finally, the boundaries distinguishing the various *FU*
173 numbers were defined. The colour transition angle α_{iT} , under which a scale number transition
174 takes place, was taken according to Eq. 6.

$$\alpha_{iT} = \frac{(\alpha_i + \alpha_{i+1})}{2} \quad (6)$$

176

177 Both α_i and α_{iT} are presented in Table 3. The *FU* numbers for a given MERIS pixel M with
178 chromaticity coordinates $x_M - x_W = -0.15$ and $y_M - y_W = 0.1$ (yellow point in Fig. 3) can be
179 determined as follows: First the angle (Eq. 5) is determined as $\alpha_M = 146^\circ$ and then is compared
180 with a simple MATLAB loop for $i=1$ to 21 twenty-one values of α_{iT} given in Table 3. From this
181 loop $\alpha_M > \alpha_{iT}$ is true for the first time reaching the angle $\alpha_{iT}=133.96$ degrees that corresponds to
182 a discrete value of *FU*=6 that is attributed to this MERIS pixel M.

183 **2.3 Ship-borne measurements**

184 North Sea and Wadden Sea (Hommersom et al., 2009) were optically sampled in 2006 (Fig. 5)
185 and several lakes and rivers were sampled in 2001, 2006 and 2007. The surface radiance L_{sfc} ,
186 sky radiance L_{sky} and incoming solar irradiance E_s were measured simultaneously, using TRIOS
187 hyper-spectral radiometers (Heuermann et al., 1999). Remote sensing reflectance was then
188 calculated as $R_{RS} = L_w/E_s$, where L_w is the water-leaving radiance ($=L_{sfc} - \rho L_{sky}$) and E_s is the

189 downward irradiance just above the sea surface (Mueller et al., 2003). To a good approximation
190 $[\rho_w]_N \approx \pi R_{RS}$ (Lee et al., 1994).

191 To illustrate the potential use of the satellite derived *FU* maps, databases containing globally
192 collected ship-borne *FU* observations were consulted. From the oceanographic and
193 meteorological database, archived by the United States National Oceanographic Data Centre
194 (NOAA-NODC; Boyer et al., 2006) and from the ocean colour database at the Royal Netherlands
195 Institute for Sea Research, *FU* observations were extracted. To create the maps, *FU*
196 measurements were interpolated through an Inverse Distance Weighted (IDW) technique
197 (Watson and Philip, 1985) in an ARCGIS environment. The IDW interpolation was carried out
198 over 2 degrees with a grid size of 0.2 degrees.

199

200 **3 Data sets**

201 The FUME algorithm was applied to five MERIS images acquired over the areas shown in Fig. 4.
202 These areas were chosen for their different sea colour properties (Wernand et al. submitted)
203 and cover the North Sea (1), the Red Sea (2, 3), Yellow Sea (4) and the Sea of Japan (5). The
204 images were extracted from ESA's on-line database, the MERIS Catalogue and Inventory
205 (MERCI, Brockman et al., 2005). MERIS Reduced Resolution (RR) geophysical products (Table 4)
206 contain, among other products, a total of 14 spectral images of normalized band reflectances
207 and the derived products for pigments (Algal-1, Algal-2, SPM and YS). A Reduced Resolution

208 image has 4×4 less pixels than the same image in Full-Resolution, thus representing an area of
209 1040m \times 1160m.

210 To validate the FUME algorithm a dataset of simultaneously collected *FU*-observations and
211 hyperspectral subsurface and above water spectra was consulted. This dataset was established
212 in 2001 and contains observations and optical data of a wide range of coloured water types, as
213 river-, lake-, coastal and open sea, with the *FU*-scale varying from *FU3* to *FU21*. From the 156
214 data sets collected (Table 2 in Hommersom et al., 2009) only one was collected close (2.5 hour
215 prior) to the MERIS image acquisition time on May 4th 2006.

216 The routine collection of *FU*-measurements in all world seas has once been very intense, mainly
217 in the 20th century, and over 220,000 measurements are known and available (Wernand et al.,
218 submitted). However, in the first decade of the 21st century the *FU* data collection is much
219 more limited and/or has not yet been recorded in the central archives. Therefore, we have
220 chosen to show data from the same seasons in earlier years: For the Red Sea, 52 observations
221 are available and were collected during the winters of 1895 to 1898. For the Yellow Sea, 2882
222 *FU* observations were collected during the winters of 1930 to 1999.

223

224 **4 Results**

225 **4.1 MERIS *FU* maps**

226 For all five MERIS images the reflectance values in band 1-9 per pixel were converted to
227 chromaticity coordinates and into a *FU* number using Eq. 3 to 5. Converted images are further

228 referred to as *FU* maps. These *FU* maps are presented in Fig. 5, Fig. 6 and Fig. 7. In these figures
229 we have used the MERIS flags per pixel to identify land (grey), clouds (white) and the failure to
230 collect observations or retrieve water-leaving reflectances (black). The legend (RGB values are
231 given in Table 5) represents the *FU* colours as close as possible. The $[\rho_w]_N$ spectral signatures at
232 the locations marked with a red circle in the maps are plotted in Fig. 8.

233 The first *FU* map shown in Fig. 5, acquisition date of 4 May 2006, covers the North Sea, the
234 Baltic and the Wadden Sea. The colour of the North Sea varies between *FU3* and *FU9*. The
235 colour within the left red circle situated between the Thames and Humber estuaries was
236 estimated as *FU6*. The central North Sea shows values of *FU3* to *FU4* with an occasionally *FU2*
237 (very blue oceanic waters). The Wadden Sea, a large intertidal sea behind multiple barrier
238 islands, north of Holland and Germany, and west of Denmark, is dominated by sediment and
239 outflow of humic-acid rich river water and has higher *FU* values, up to *FU=18*.

240 Figure 6 shows a winter *FU* map of both northern- and southern Red Sea taken on 22 and 23
241 December 2003, respectively. The colour of the northern Red Sea is mainly *FU2* to *FU3* with
242 maximum values of *FU5*. The southern part, which is shallower than the northern part, shows a
243 possible plankton bloom starting south of 17.5°N with values of *FU8* (red circle) to *FU11*. The
244 water flowing through the narrow strait of Bab-al-Mandab into the Gulf of Aden (area at the
245 most south-eastern point on the map) shows much bluer values: *FU2* to *FU4*.

246 Figure 7 shows a *FU* map of the Yellow Sea (left) and the Sea of Japan (right) acquired on 11
247 February 2009 and 14 June 2004, respectively. The outflow of the Yangtze River (south of the
248 red circle in the left image) shows high *FU* values, between *FU7* up to real brownish colours of

249 *FU19*. Unfortunately, the area close to the river outflow is flagged as “No Data area”. Within the
250 red circle a value of *FU9* is calculated. The Sea of Japan (right panel) shows values of *FU2* to *FU3*
251 (within the red circle the value is *FU2*). Remarkable is the relative green area east of Hokkaido
252 (*FU9* marked by the red circle east) with values up to *FU10*. To verify our results the MERIS
253 level2 chlorophyll product was consulted, which showed high concentrations of chlorophyll-a
254 ($>2\text{mg}\cdot\text{m}^{-3}$) East of Hokkaido and concentrations between 0.1 and $0.5\text{mg}\cdot\text{m}^{-3}$ in the Sea of
255 Japan.

256

257 **4.2 Ground truth**

258 The reflection spectrum at the match-up station in the Wadden Sea (WS-GT) is plotted in Fig. 8
259 and appears very similar in shape to the MERIS spectrum (WS). By extraction of the reflection at
260 exactly all 9 MERIS bands and running the FUME algorithm a value of $FU=15$ was retrieved. The
261 MERIS pixel at this location (red circle in the Wadden Sea) has a calculated *FU* value of 14,
262 which is in good agreement with the ground truth *FU* value concerning possible adjacency
263 effects of tidal flats within the pixel.

264 The MERIS water-quality products and FUME results were extracted along a transect (yellow
265 line in Fig. 5) perpendicular to the coast. The transect starts at the Match-up point in the
266 Wadden Sea (red circle) and ends in the Central North Sea. The results are shown in the three
267 panels of Fig. 9. Within the first 30 pixels the waters are within or close to the Wadden Sea,
268 characterised by very high loads of sediment ($> 1 \text{ g m}^{-3}$) and Yellow Substance (absorption > 1
269 m^{-1} at 442 nm), corresponding to *FU* values above 7. In the next part (pixels 30-170), the *FU*-

270 values show a gradual gradient from 7 to 3, reflecting a decrease in algal pigments (both Algal-1
271 and Algal-2 products) because both YS and SPM are rather constant in this interval. An
272 interesting feature can be observed in the pixels (170-300) where the Case-2 water algorithm
273 seems to fail (unrealistic high SPM and YS), likely due to additional scatter by cirrus clouds.
274 Fortunately, the *FU*-scale seems robust and corresponds rather well with the algal-1 product.
275 Based on field measurements, a comparison between observed and modelled *FU* numbers was
276 made (Fig. 10). The correlation between observed and modelled *FU* numbers is around the 1:1
277 line. However, full spectral derived *FU* correlates better ($R^2 = 0.81$) with in-situ data than MERIS
278 derived *FU* ($R^2 = 0.74$). More spectral information provides a more reliable *FU* estimate. The
279 largest outliers were found in the 11-17 *FU* range. These are the green-yellowish water colours,
280 which corresponds to ~ 500 -600 nm of the visible light. This is also the part of the light
281 spectrum with is covered by only one MERIS band (560 nm) which leads to the largest loss of
282 information.

283 Comparing the MERIS winter *FU* maps of the northern and southern Red Sea of Fig. 6 with the
284 winter IDW *FU* map of Fig. 11 similar patterns can be recognized despite a time cap of over a
285 century between data acquisition. When we compare these maps it seems that the colour of
286 the Red Sea did not change significantly over time, although we cannot say anything about
287 intermediate colour changes between 1899 and 2002. Within the red circles the MERIS *FU* map
288 (Fig. 6) shows *FU2* for the northern location and *FU8* for the southern location, while the IDW
289 *FU* map gives identical values at both locations.

290 The *FU* map for the Yellow Sea, based on 2882 *FU in-situ* observations collected during the
291 winters of 1930 to 1999, is shown in Fig. 12. The Yellow Sea shows *FU* numbers of *FU4* at open
292 sea areas to values of *FU20* in front of the outflow of the Yangtze River. Both, the MERIS map of
293 Fig. 7 and the IDW interpolation of Fig. 12 show similar colour patterns. The red circles in the
294 MERIS map indicate *FU9* in the lower and *FU11* in the upper red circle. In the IDW map
295 respectively *FU12* and *FU14* are indicated. A possible explanation of the bluing of the water
296 (bluer colours show up in the 2008/9 map) is the reduced outflow of Yangtze water into the
297 Yellow Sea due to the hydroelectric Three Gorges Dam which became operational in 2003. The
298 effect is a reduced upwelling and thus productivity resulting in less green water (Chen, 2000;
299 Gong et al., 2006).

300

301 **5 Discussion and conclusions**

302 In this paper an algorithm is presented that allows retrieval of the Forel-Ule sea colour from the
303 MERIS satellite sensor. The Forel-Ule colour can be seen as the colour standard closest to the
304 real colour of water. The elegance of our algorithm is that it converts multispectral
305 observations to one simple number that is only dependent on a well known universal set of
306 colourimetric functions. The classification of sea water is simplified by means of a numerical
307 value between 1 and 21, instead of a classification by a normalized water-leaving spectral
308 reflectance signature, or the concentrations of the dominant optical constituents.

309 The approach is demonstrated by the processing of multispectral observations of oceans and
310 coastal waters made by the MERIS ocean colour sensor, to *FU* maps that cover colour classes

311 between indigo blue, green and brown. Five different seas were selected worldwide; these
312 were processed to obtain *FU* maps. The maps show very detailed patterns and gradients,
313 mainly in the near coastal zone as expected by the more outspoken hydrographical gradients
314 there. When the MERIS maps of sea and ocean colour distribution were compared with ground
315 truth Forel-Ule observations mapped in the same season, similar patterns and *FU* numbers
316 were observed, even when *FU*'s of more than a century ago were processed. This opens new
317 ways to study the spatial and temporal evolution of the colour of the sea worldwide. The FUME
318 algorithm can easily be adapted to data from other satellites which have enough bands in the
319 visible part of the spectrum to properly derive the colour of the water.

320

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329

330

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