

Dear Editor,

We are glad to resubmit our manuscript for consideration to publication in ESSD. Reception by the reviewers was positive and only minor changes were suggested. We addressed them in the revised version attached. Most important change is the decision to share the code to generate the hue angle, the Forel-Ule index and the Secchi disk depth from satellite data, through the upload to a public repository that is linked in the revised version. We additionally screened the text and improved the wording here and there to further improve its understanding. In order to be fully transparent, in addition to the comment-by-comment reply to reviewers, we also attach the manuscript with all changes highlighted.

Best regards,

Jaime Pitarch.

Reply to anonymous Referee #1

The authors present a publicly available, monthly, global dataset of the Forel-Ule index, hue angle, and Secchi disk depth derived from the ESA Ocean Colour Climate Change Initiative satellite times series of spectral remote-sensing reflectances. Generally speaking, the manuscript is well written and straightforward – almost terse, which I find acceptable as the algorithms and approaches are well documented. Overall, I see no major show-stoppers in its publication. Minor comments are provided below, most of which focus on grammar and presentation.

We appreciate the positive opinion of the referee and thank him/her very much for the time and feedback to improve the manuscript. We have taken all comments in consideration and have improved the manuscript accordingly. We agree on the “terse style”, but unfortunately ESSD articles can be like this, as the journal forbids any scientific interpretation of the data. Some minor editing has been made here and there to make parts smoother to read. With the example of the North Atlantic Ocean, we intended to provide some hints on further use and make the data more interesting for scientists.

Line 38: Suggest expanding satellite acronyms, adding agencies, and adding mission timelines, e.g., “the NASA Moderate Resolution Imaging Spectroradiometer onboard Aqua (Aqua-MODIS; 2002-present)” – or, changing “present” in the former to the last year included in OC-CCI.

We have expanded the sensors’ names, though not been explicit on the carrying platforms, in order to not to make the text cumbersome.

L42: Please indicate spectral dependency and provide units the first time a variable is introduced.

Done.

L72: Remove out of place “Rrs” or revise to “calculate them from Rrs”.

Corrected.

L80: “Downloaded product is the” is grammatically awkward. Do you mean “The source product for all derived variables presented in this dataset...”.

Indeed, this is much better. Corrected.

L99: Expand CIE acronym.

Done.

General comment on uncertainties: Just confirming that temporal, pixel-level uncertainties are not possible? This is generally apparent from this section, but would still be useful for clarity to state as such up front.

Indeed, the estimation of pixel-level uncertainties is not possible for the products released in this dataset. We are aware of this lack, but such information can only be given after a “spectral class discrimination”, using a large and quality-controlled in-situ dataset that spans the broadest dynamic range. In fact, such stringent requirements have the effect that not all products can be released with pixel-wise uncertainties in the OC-CCI dataset (Jackson et al., 2019). We have elaborated on this point in section 4. Validation in the revised version.

L135: Suggest mentioning that any systematic bias in the geophysical products will not be reduced like random noise.

Done.

L154: What is the remaining uncertainty estimate after 7.5% was subtracted in quadrature? Or, was the subtraction included in the “within 20%” stated on line 149? Even if reported elsewhere, the spectral values of calculated Rrs uncertainties (used on line 57) are worth repeating here. What other terms are included in the (quadrature-expressed) uncertainty budget?

The 7.5 % is a reasonable uncertainty estimate of the in-situ Rrs measurements. The statistics “within 20 % ” were taken from Table 3 of Pitarch et al. (2020) where we made a matchup between in-situ Rrs and OC-CCI Rrs. Subtraction in quadrature of 7.5 % from these Rrs uncertainties are the input to the Monte Carlo simulations of the hue angle, Forel-Ule and the Secchi disk depth whose results we report in Figure 1 and Table 2 of the present paper. All in all, we think that we have provided sufficient information for readers to follow and even duplicate out Monte Carlo experiment. Note that, following the suggestion by Reviewer 2, the code has been made publicly available.

Line 252: Global “patterns”. And, suggest rewording to “Despite the scale ranging from 0 to 21”.

Corrected.

Line 264: Neither the 0.25-deg nor the 1-deg datasets have been described yet. Generation of these datasets needs to be presented earlier in the manuscript.

These datasets are now introduced in section “3. Product description”.

References

Jackson, T., Chuprin, A., Sathyendranath, S., Grant, M., Zühlke, M., Dingle, J., Storm, T., Boettcher, M., and Fomferra, N.: Ocean Colour Climate Change Initiative (OC_CCI) –Interim Phase. Product User Guide. D3.4 PUG, 2019.

Pitarch, J., Bellacicco, M., Organelli, E., Volpe, G., Colella, S., Vellucci, V., and Marullo, S.: Retrieval of Particulate Backscattering Using Field and Satellite Radiometry: Assessment of the QAA Algorithm, *Remote Sensing*, 12, 77, <https://doi.org/10.3390/rs12010077>, 2020.

Reply to anonymous Referee #2

Interactive comment on “Global maps of Forel-Ule index, hue angle and Secchi disk depth derived from twenty-one years of monthly ESA-OC-CCI data” by Jaime Pitarch et al.

This study provides a new, open access dataset that consists of global maps of the Forel-Ule index, hue angle and Secchi disk depth and can be conveniently downloaded from PANGAEA. A merged multi-sensor data (OC-CCI) was used as the source data and the algorithms can be traced from other documents. Generally, this is meaningful work and facilitates the research of other scientists in the water color remote sensing community. I suggest making the following minor revisions before the publication of this study:

We greatly appreciate the reviewer’s positive view on this article and acknowledge the careful reading and the useful suggestions to improve the text. We have made the following changes:

L26: “easier to handle” should be “easier-to-handle”

Corrected.

L49: delete “the” before “water surface”

We believe the presence of the “the” is necessary, because the action of lowering is with respect to the surface.

delete “is” before “tracked”

Corrected.

L50: delete “,” before “(Wernand, 2010)”

Corrected.

L72: better to change “so far” to “thus far”

Corrected.

“them” should be “their”

We believe “them” is the appropriate word here, because it does not refer to persons.

L95: delete “the” before “deep blue”

Corrected.

L133: As daily OC-CCI products are also available and can be used to match with in-situ data, why are daily variables not included in this new dataset? Or as an alternative, if possible, the authors could publish their code on GitHub, perhaps a function that makes Rrs the input and FU index and other variables the output.

Thank you, this is a great suggestion. We have uploaded the code that calculates the hue angle, Forel-Ule index and Secchi disk depth from Rrs to a GitLab repository, with the appropriate link given in the article.

We focused here on monthly data because that is the time resolution commonly used in satellite climate studies. Daily data will be much more demanding in terms of processing and storage, but we may consider such a release if the present dataset has a good reception.

L143: delete “and” before “without”

Corrected.

L155: The minimum “exact” z_{SD} in these three experiments is set to 8.0 m, which limits the verification to case 1 waters. But it is obvious that the nearshore seawater will be much more turbid; therefore, is the dataset provided by this research still reliable in turbid water (for example $FU > 10$)?

Indeed, this algorithm is applicable to waters with z_{SD} values of less than 1 m, as is evidenced by the good match between the in-situ z_{SD} and the Rrs -derived z_{SD} ; see Lee et al. (2015). The choice of a particular spectrum that leads to $z_{SD} = 8.0$ m is just one arbitrary example among three to illustrate the effect of uncertainty in Rrs on the derived z_{SD} value.

Table 2: The “exact value” of FU index in EX.2 does not match that in Fig. 1, please check it. And please change “ $A(\circ)$ ” to “ $\alpha(\circ)$ ” to keep it consistent with Fig. 1.

Corrected.

L189: “RMS=22.8%”, do you mean “relative RMS=22.8%”?

Yes, these are relative units.

L195: How is this “~32%” calculated, “22.8%”+“10%”?

Uncertainties are added in quadrature. This is now explicitly mentioned in sections 4.1 and 4.2 .

L290: “variation” should be “variations”

Corrected.

Global maps of Forel-Ule index, hue angle and Secchi disk depth derived from twenty-one years of monthly ESA-OC-CCI data

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Abstract. We document the development and public release of a new dataset (1997-2018), consisting of global maps of the Forel-Ule index, hue angle and Secchi disk depth. Source data comes from the European Space Agency (ESA) Ocean Colour (OC) Climate Change Initiative (CCI), which is providing merged multi-sensor data from the mid-resolution sensors in operation at a specific time from 1997 to the present day. Multi-sensor satellite datasets are advantageous tools for ecological studies because they increase the probabilities of cloud-free data over a given region, as data from multiple satellites whose overpass times differ by a few hours are combined. Moreover, data merging from heritage and present satellites can expand the duration of the time series indefinitely, which allows the calculation of significant trends. Additionally, data are remapped consistently and analysis-ready for scientists. Also, the products described in this article have the exclusive advantage of being linkable to in-situ historic observations and thus enabling the construction of very long time series. Monthly data are presented at a spatial resolution of ~4 km at the equator and are available at PANGAEA, <https://doi:10.1594/PANGAEA.904266> (Pitarch et al., 2019a). Two smaller and ~~easier-easier-to-to~~ handle test datasets have been produced from the former: a global dataset at 1 degree spatial resolution and another one for the North Atlantic at 0.25 degree resolution. The computer code for the generation the Forel-Ule index, hue angle and Secchi disk depth from a given remote-sensing reflectance is also shared at <https://gitlab.com/jaipipor/hue-fu-secchi> and can be easily set in loop mode for batch calculations.

1. Introduction

Single-sensor satellite datasets are not long enough to provide significant evidence of climatic trends that become manifest over fluctuations, so the multi-sensor approach has been fostered since the last decade (IOCCG, 2007). In Europe, the ESA's Climate Change Initiative (CCI) was created to address the United Nations Framework Convention on Climate Change requirements on a systematic monitoring ~~of-of~~ the global climate system. The Global Climate Observing System (GCOS) defined a number of Essential Climate Variables (ECVs), which are

physical, chemical or biological variables that critically contribute to the characterisation of Earth's climate. Ocean Colour is one of the ECV and the OC-CCI is devoted to producing a time series of consistent measurements encompassing water-leaving radiance in the visible domain, derived chlorophyll and inherent optical properties, based on merged [data from the MEdium Resolution Imaging Spectrometer \(MERIS\) \(2002-2012\)](#), [MERIS](#), [the Moderate Resolution Imaging Spectroradiometer \(MODIS\) Aqua-MODIS\(2002-present\)](#), [the Sea-Viewing Wide Field-of-View Sensor \(SeaWiFS\) \(1997-2010\)](#) and [the Visible Infrared Imaging Radiometer Suite \(VIIRS\)VIIRS \(2012-present\) satellite sensors data](#) (Jackson et al., 2019).

While the chlorophyll concentration has historically been the most studied variable from ocean colour observations, more variables can be derived with confidence that can provide further insights, such as inherent optical properties or the diffuse attenuation coefficient ($K_d(\lambda); m^{-1}$). Algorithm developers can also download the remote-sensing reflectance ($R_{rs}(\lambda); sr^{-1}$) at <ftp.oceancolor.org>, and derive any other geophysical property themselves.

In the last decade, two historic optical variables have received renewed attention: the Secchi disk depth ($z_{SD}; m$) and the Forel-Ule (FU; [unitless](#)) colour index. These two are the oldest oceanographic variables that are directly related to ocean colour, recorded for many decades before the advent of satellites (Boyce et al., 2012) and publicly available globally (NOAA, 2013). The Secchi disk is a [simple](#)-white disk that is lowered from above the water surface and [is](#)-tracked visually until it goes out of sight. The depth when it ceases to be visible provides a reading of water transparency, (Wernand, 2010). z_{SD} is influenced by the underwater light attenuation, which in turn depends on the light absorption and scattering caused by the varying concentrations of dissolved and suspended substances in the water. The z_{SD} observation also depends on [the](#) disk design and environmental factors that need to be understood (Pitarch, 2020). The popularity and simplicity of this practice allowed its generalized use in oceanic surveys during the last century, although its use [significantly](#) declined [significantly](#) in the last decades. Currently, the method is experiencing a revival due to recent efforts to derive this quantity from remote sensing data (Lee et al., 2015).

The FU scale was developed as a set of standard colours that would allow visual water colour indexing (Wernand and Van Der Woerd, 2010). As with z_{SD} , the colour of natural waters is also influenced by [the](#) substances in the water, present in different concentrations, which constitutes the basis for ocean colour science. The popularity of the FU scale has had a similar history as the Secchi disk, and has evolved from being widely adopted, then mostly abandoned and recently revived, aided by the link to remote-sensing data (Wernand et al., 2013).

The hue angle is [a popular quantity in the field of image colour analysis](#). It is used as an intermediate quantity in the process of deriving FU from any light spectrum (Wernand et al., 2013) and it can be regarded as the continuous companion of the FU scale. Specific algorithms have been developed for the spectral characteristics of various ocean colour sensors (Van Der Woerd and Wernand, 2015; Van Der Woerd and Wernand, 2018). In terms of in-situ sampling, this quantity cannot be estimated visually but only radiometrically, although [Red-Red-Green-Green-Blue \(RGB\)](#) images have proven good for this sake, thus enabling the use of digital cameras [for citizen science](#) (Ceccaroni et al., 2020).

Pitarch et al. (2019b) made an analysis of global seasonal variability of these parameters using the climatological ESA-OC-CCI data, including some cross-relationships and ranges of variability.

75 Despite the demonstrated usefulness of these optical parameters in marine physics and biogeochemistry, no public dataset of them exists ~~so thus~~ far, so scientists need to calculate them ~~from~~ R_{rs} privately, which, other than creating redundant efforts, may be ~~a too high wall to overcome discouraging~~ for non-experts. This gap is filled in the present article with the provision of the monthly series from 1997 to 2018 of Secchi disk, hue angle and Forel-Ule index at ~4 km resolution.

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2. Methods

2.1. Ingested data

85 ~~The source product for all derived variables presented in this dataset~~ ~~Downloaded product~~ is the merged multi-sensor OC-CCI v4.2 ~~remote sensing reflectance~~ (R_{rs}), which is the primary quantity used for ocean colour studies, containing only spectral information related to the colour of the water, while having all interfering factors ~~ideally~~ removed in the atmospheric correction and normalization calculations. R_{rs} is provided at a monthly frequency and projected on a rectangular grid of 2.5 minutes of arc for both latitude and longitude, which corresponds to about 4 km at the equator, and decreasing poleward. For more specific details, one can read the Product User Guide (Jackson et al., 2019). ~~Every file of Secchi disk, hue angle and Forel-Ule keeps the same format as the~~ 90 ~~corresponding source R_{rs} file.~~ Numeric precision is single (32-bit floating-point values), except time, which is 32-bit signed integer. Every downloaded data file contains a map of estimated bias in R_{rs} , which we add to generate the unbiased R_{rs} estimate. This bias compensation, which users need to perform themselves, has proven to enhance the quality of retrievals of R_{rs} -derived geophysical products like optical particle backscattering (Pitarch et al., 2020). ~~Every generated file of Secchi disk, hue angle and Forel-Ule keeps the same format as the~~ 95 ~~corresponding source R_{rs} file.~~

2.2. The Forel-Ule and hue angle algorithms

100 The hue angle (α ; θ) is a continuous numerical variable that expresses the colour of a light spectrum ~~with-in~~ a single number. In natural waters, ~~the hue angle~~ ranges from about 40° in brown waters to about 235° in the deep blue oceanic waters. As the ~~α definition involves three spectral integrals across source spectra and colour perception cover~~ the continuous spectral range, some kind of adaptation needs to be done when applying it to satellite data that is provided only at a small number of wavelengths. Full technical details about ~~α hue angle~~ determination from ~~multispectral~~ R_{rs} are published (Van Der Woerd and Wernand, 2015; Van Der Woerd and Wernand, 2018; Pitarch et al., 2019b).

105 The FU scale was optically characterized as a canonical set of points (x, y) in the [Commission Internationale de l'Éclairage \(CIE\)](#) space (Novoa et al., 2013). Because ~~α the hue angle~~ is the polar angle of the (x, y) coordinates, the ~~α hue angle~~ of all canonical FU points can be calculated. Thus, given a each R_{rs} spectrum, its α can be calculated and ~~subsequently, any measured hue angle can be assigned~~ an FU index ~~can be assigned it~~ by choosing the nearest FU index in ~~hue angle α unit~~ terms. In fact, FU can be seen as a discrete version of ~~α the hue angle~~ and can be used to cluster water masses in terms of their colour. However, it was found that the FU scale lacks resolution and dynamic range for the most oligotrophic oceanic waters (Pitarch et al., 2019b). Vast regions of the ocean remain saturated all year round at FU=1. For this reason, a new lower end FU=0 was proposed in order to introduce more differentiation within the bluest waters. This dataset includes that addition.

2.3. The Secchi disk depth algorithm

115 We apply the state-of-the-art algorithm for the remote estimation of the Secchi disk depth (z_{SD}) by Lee et al. (2015). According to the underlying theory, z_{SD} essentially depends on the spectral minimum of the diffuse attenuation coefficient (K_d), which is analytically modelled as a function of the absorption (a) and optical backscattering coefficient (b_b) (Lee et al., 2013). The two latter are retrieved with the QAAv6 algorithm (~~Lee et al., 2002~~). ~~(Lee et al., 2002), including a~~ Notably, the retrieved b_b is empirically corrected for Raman scattering after Lee et al. (2013). It has been shown that such step improves the match of the retrieved b_b against reference data (Pitarch et al., 2020). Solar zenith angle, which is also needed for K_d , is set equal to zero, to be consistent with the reflectance normalization convention used for in the OC-CCI dataset satellite normalized reflectances.

3. Product description

125 The processing is applied on a pixel basis from every source netCDF R_{rs} file. A sample code that applies the former algorithms to generate the hue angle, Forel-Ule index and Secchi disk depth from a given R_{rs} can be downloaded from <https://gitlab.com/jaipipor/hue-fu-secchi>. This code can be easily set in loop mode and parallelized, as well as linked to a netCDF file generator code, which is how the files of this dataset were produced.

~~Therefore, t~~The format of the product file is the same and the spatial and temporal variables are directly copied from the source file. The hue angle, the Forel-Ule index and the Secchi disk depth are added. Content of each file is summarized in Table 1. Every netCDF file has the naming ESACCI-OC-L3S-MERGED-130 1M_MONTHLY_4km_GEO_PML_Hue_FU_SD-yyyy-mm-fv4.0.nc, where yyyy is a four character string for the year and mm is a four character string for the month.

From this product, two smaller, reduced-resolution products, were generated: (1) a global dataset at 1° spatial resolution, more than enough to detect spatial features for global analyses, and (2) a North Atlantic (0 - 60° N, - 80 - 0° E) dataset at 0.25° resolution, that can be used to study regional trends in this region at an increased resolution yet keeping the computational cost reasonably low.

Table 1 File description

Parameter	Description	Dimensions	Size (pixels)
lat	Geographic latitude, positive northward	degree	4320x1
lon	Geographic longitude, positive eastward	degree	8640x1
time	Time since 1970-01-01 00:00:00	day	1x1
alfa	Hue angle, according to (Pitarch et al. (2019b))	degree	8640x4320x1
FU	Extended Forel-Ule index of the remote-sensing reflectance (0 to 21), according to (Pitarch et al. (2019b))	-	8640x4320x1
z_SD	Secchi disk depth, according to Lee et al. (2015)	m	8640x4320x1

4. Validation

Validation of water quality parameters is an ongoing activity, even for consolidated products like chlorophyll and reflectance. In the remote sensing community, validation is mostly known as a comparison to in-situ reference data. Assuming errorless reference data, uncertainties are then due to the atmospheric correction and the applied water quality algorithms, though. However, in-situ data also contains uncertainties that are seldom considered.

Monthly OC-CCI products cannot be used for matchup to in-situ data as matchups are intended for instantaneous data, with time coincidence between the satellite and the in-situ readings recommended to be within few hours. Understanding this rule requirement in a broad sense, daily satellite data, also provided within OC-CCI, may be acceptable. Therefore, the uncertainties can be assessed using daily data, knowing that uncertainties present in the monthly aggregates will be smaller, as the temporal averaging will dampen random noise by a factor equal to the square root of the number of observations at every pixel for every month, and, in fact, this can be seen in the produced monthly images, where speckle noise looks rather low. Unfortunately, this number of observations per pixel in each month is unknown for the monthly OC-CCI products, reduced noise. This can be seen in the produced monthly images, where speckle noise looks rather low. Systematic biases will remain after averaging, though.

Ideally, uncertainties would be provided at pixel level for every product map, but such information could only be given after an uncertainty differentiation across the product dynamic range, using a large and high quality in situ dataset that spans the broadest dynamic range, followed by a clustering into spectral classes (Jackson et al., 2019). In fact, such a stringent requirement makes that, for some products like backscattering, uncertainties cannot be provided as maps. As the situation for the variables treated in this article is similar in terms of paucity of publicly of quality-controlled available data, uncertainties will be provided as bulk estimates.

4.1. Remote-sensing reflectance

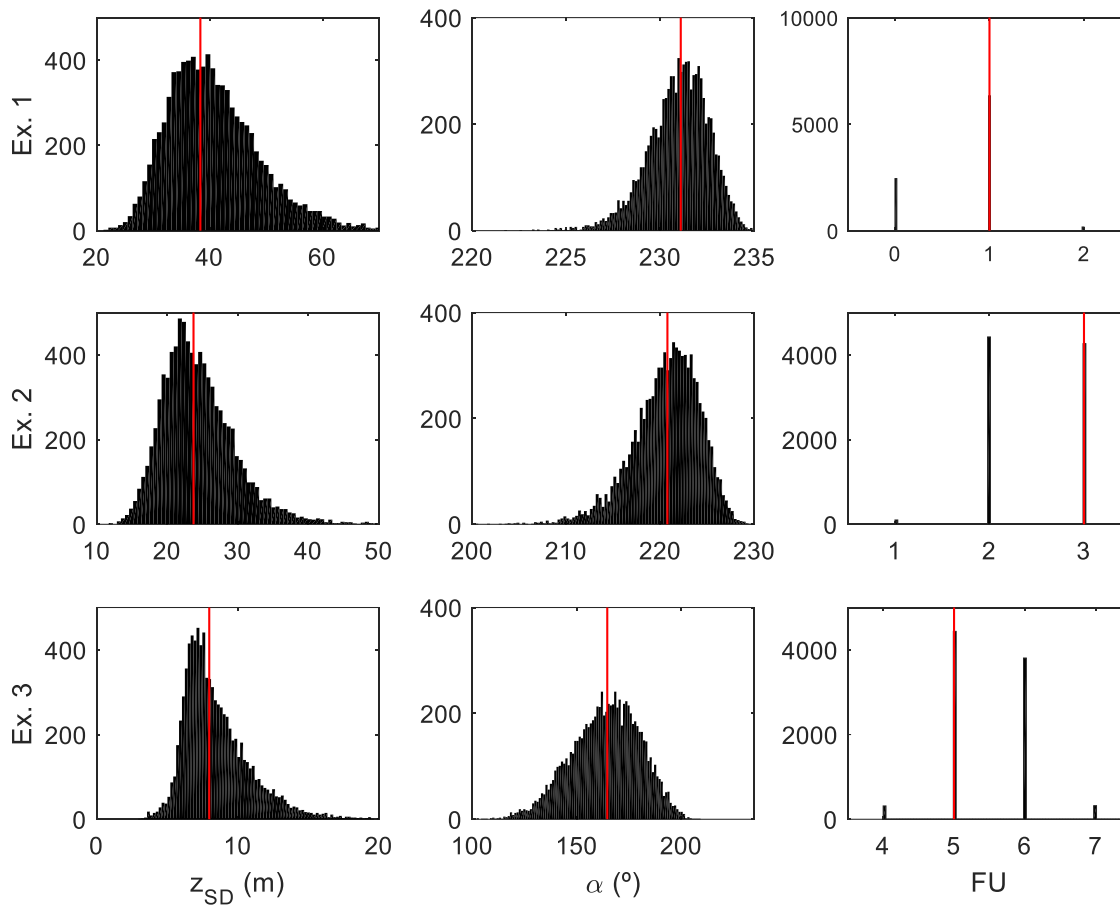
The OC-CCI's Product User Guide (PUG) (Jackson et al., 2019) documents uncertainty assessment of satellite OC-CCI R_{rs} by comparing them to a large in-situ R_{rs} database that merges in-situ R_{rs} from many different sources with very disparate instrumentation and quality standards (Valente et al., 2019) and without an uncertainty budget assessment. Usage of this entire database can artificially increase the error statistics when compared to satellite data. For this reason, in a previous publication (Pitarch et al., 2020; Table 3) we compared daily OC-CCI reflectances to Valente's database as well as to other two internal databases. Uncertainties were the lowest when OC-CCI daily products were compared to an internal database "CNR", collected using strict NASA protocols and

uncertainties traceable to NIST standards. Relative RMS errors were within 20 %-, with biases not significantly different from zero, except for the band at 412 nm. At 670 nm, errors were much higher in relative terms due to the low value of $R_{rs}(670)$, but they remained low in absolute units (sr^{-1}). These RMS errors include uncertainties present in the in-situ R_{rs} as well, which may range between 5% to 10% (Zibordi et al., 2019), so an educated guess of a spectral average of 7.5 % was subtracted in quadrature from them to isolate the errors in the satellite data.

These remaining uncertainty estimates need to be propagated by the Secchi disk depth, hue angle and Forel-Ule algorithms. To assess this matter, three example R_{rs} spectra were taken, representing ultra-oligotrophic, oligotrophic and eutrophic waters. Their respective Secchi disk depth, hue angle and Forel-Ule were calculated. Posteriorly, a Monte-Carlo analysis was made by adding the R_{rs} uncertainties as Gaussian independent noise. This was repeated 10000 times and the histograms (Fig. 1) and related statistics (Table 2) of the resulting Secchi disk depth, hue angle and Forel-Ule were retrieved. It is shown that, for Secchi disk, biases have a spread of about $\sigma=22\%$ around the errorless value, with a negligible mean bias, increasing towards more turbid waters. The hue angle experiences a spread of about $\sigma=1\%$, also with a negligible bias, though this result has to be interpreted keeping in mind the high average value and the low dynamic range of the hue angle. Statistics of the FU are much harder to interpret due to the strongly discrete nature of this variable, but essentially FU oscillates between the central value and the adjacent ones, in proportions that depend on the distances to these in hue angle terms.

A third uncertainty source is the remote algorithm itself with respect to the actual in-situ value. Since the algorithms were published elsewhere, this article does not validate them further. Here, we simply summarize statistical findings on their accuracy in each of the three next sub-sections.

185



190 Fig. 1 Monte Carlo simulation of error propagation from **OC-CCI spectrum R_{rs} remote sensing reflectances** to Secchi disk depth, hue angle and Forel-Ule index. The upper panel shows the results of 10000 simulations (black bars) for ultra-oligotrophic waters, compared to the reference spectrum results (red line). The middle panels show the results for an oligotrophic water spectrum and the lower panels depict the results for a eutrophic water spectrum.

195 Table 2 Biases and standard deviation (σ) of the errors caused by random additive Gaussian perturbations to three given remote sensing reflectances from which the Secchi disk depth, the hue angle and the Forel-Ule index are derived. The “exact values” correspond to the red vertical lines in Fig. 1.

z_{SD} (m)			α (°)			FU		
Exact value	Bias (%)	σ (%)	Exact value	Bias (%)	σ (%)	Exact value	Bias (%)	σ (%)
38.4	6.9	22.5	231	-0.06	0.7	1	-25.4%	47.7%
28.5	3.3	22.7	226	-0.05	1.6	2.3	-17.1%	18.0%
8.0	6.6	30.3	165	-0.5	10.0	5	9.3%	12.6%

4.2. Secchi disk

The Secchi disk depth algorithm was validated by Lee et al. (2015). The z_{sd} derived from in-situ R_{rs} compared well to actual measurements with a Secchi disk, with a mean absolute percent difference (MAPD) of ~19 %. The linear regression slope between in-situ and remote estimation deviated marginally from the 1:1 line. Unfortunately, most of their data was not open to external users and the MAPD is a statistical parameter different from RMS or standard deviation. However, a quick numerical verification with Gaussian noise (not shown) reveals that the ratio between the relative RMS error and MAPD for unbiased estimates is very stable at about ~1.25. Therefore, Lee's MAPD=18.2 % equals to RMS=22.8 %. From this uncertainty, one must subtract the part due to the in-situ radiometry. Unfortunately, in Lee's data, radiometric uncertainties are not traced, so a reasonable educated average 10 % RMS error is guessed here. Subtracted in quadrature from the 22.8 %, results in an uncertainty of ~20 % for Secchi disk model estimations respect to in-situ data. Overall, ~~taking into consideration~~ adding in quadrature the uncertainties due to the satellite radiometry, we may conclude that Secchi disk depth derived from daily OC-CCI data has an uncertainty of ~32 % when compared to in-situ data.

4.3. Hue angle

Uncertainties in the ~~α hue angle~~ product derive from the approximation of the spectral integrals involved in the algorithm from a limited set (6 here) of satellite bands (Van Der Woerd and Wernand, 2015; Van Der Woerd and Wernand, 2018; Pitarch et al., 2019b). These uncertainties vary with the water type because the few available spectral bands may be enough to capture all the optical information in simple case 1 waters, but not in optically complex waters. In fact, for case 1 waters, in relative terms, this amounts to less than 0.1 %, while in optically complex waters, it can amount up to a 3 %. We can therefore conclude that, for satellite ~~α hue angle~~ estimations, the uncertainty is almost totally caused by the satellite radiometry, which ranges from less than 1 % to about 11 % as water shifts from very clear to turbid. This may also be an upper bound because we have assumed spectrally independent Gaussian noise at every satellite band, but errors of an atmospheric correction may have some spectral correlation. As a consequence of this, satellite-derived hue angle (and FU) seem to be particularly robust to uncertainties in satellite R_{rs} after atmospheric correction (Wang et al., 2018).

4.4. Forel-Ule index

In-situ FU readings are different depending on whether a submerged Secchi disk is used to obtain a visual estimate of the colour over its surface or the estimate is obtained by looking at the water only. Essentially, an underwater white object reflects light that, after propagating across the water column back, is observed from above with a hue angle that is lower than that of the surrounding water. These differences are significant and the amount of this bias depends on the ~~color~~ colour of the water itself effect, which was previously shown after radiative transfer simulations (Pitarch, 2017), and recently shown experimentally (Nie et al., 2020). In addition, in-situ FU can be either estimated visually or derived mathematically from a R_{rs} that has been measured in-situ. The equivalence of both must not be taken for granted and their relation has not been studied yet. In summary, there are several methodological concerns that affect the definition of reference FU measurements for comparison to satellite estimations.

235 In the interim, we choose to define reference FU measurements as those originating from in-situ hyperspectral R_{rs} , of which the hue angle has been calculated by integration and FU by discretization. In such case, the uncertainties are only those related to the satellite radiometry, as with the hue angle, with the overwhelming number of times, an error within ± 1 .

5. Global average maps

240 In order to provide the reader with a feeling of the global variability covered by these parameters, some basic statistics are demonstrated in this section. These results were produced using the reduced resolution global product at 1° . At every pixel, the 256 values (21 years and 4 months of data) were combined and the arithmetic mean, standard deviation and coefficient of variation were calculated. Fig. 2 shows the results for the z_{SD} time series. The mean map shows a global pattern that is similar when deriving other parameters like the surface chlorophyll-a concentration. In the oligotrophic gyres, there is the chlorophyll concentration is at minimum values ($<0.1 \text{ mg m}^{-3}$) in agreement with the expected higher transparency induced by low concentration of phytoplankton cells, as also highlighted by the z_{SD} . For example, The mean map highlights the oligotrophic gyres, with the South-Pacific subtropical gyre one reaching a mean of ~ 60 m in its most transparent core. ~~The map shows a global pattern that is similar when deriving other parameters like chlorophyll. In the oligotrophic gyres, the chlorophyll concentration is at minimum values ($<0.1 \text{ mg m}^{-3}$) in agreement with the expected higher transparency induced by low concentration of phytoplankton cells, as also highlighted by the z_{SD} .~~
245 ~~The standard deviation of z_{SD} shows interesting patterns. It is mostly driven by the seasonal variability and it is highest in the zones of the gyres that are furthest apart from the equator boundaries, showing the seasonal expanding and shrinking of the subtropical gyres though they might also include the long-term effect of gyre expansion (Polovina et al., 2008). Other peaks of variability are zones affected by seasonal blooms like off Iceland or the Antarctica coast. This latter feature is most accentuated in the coefficient of variation.~~
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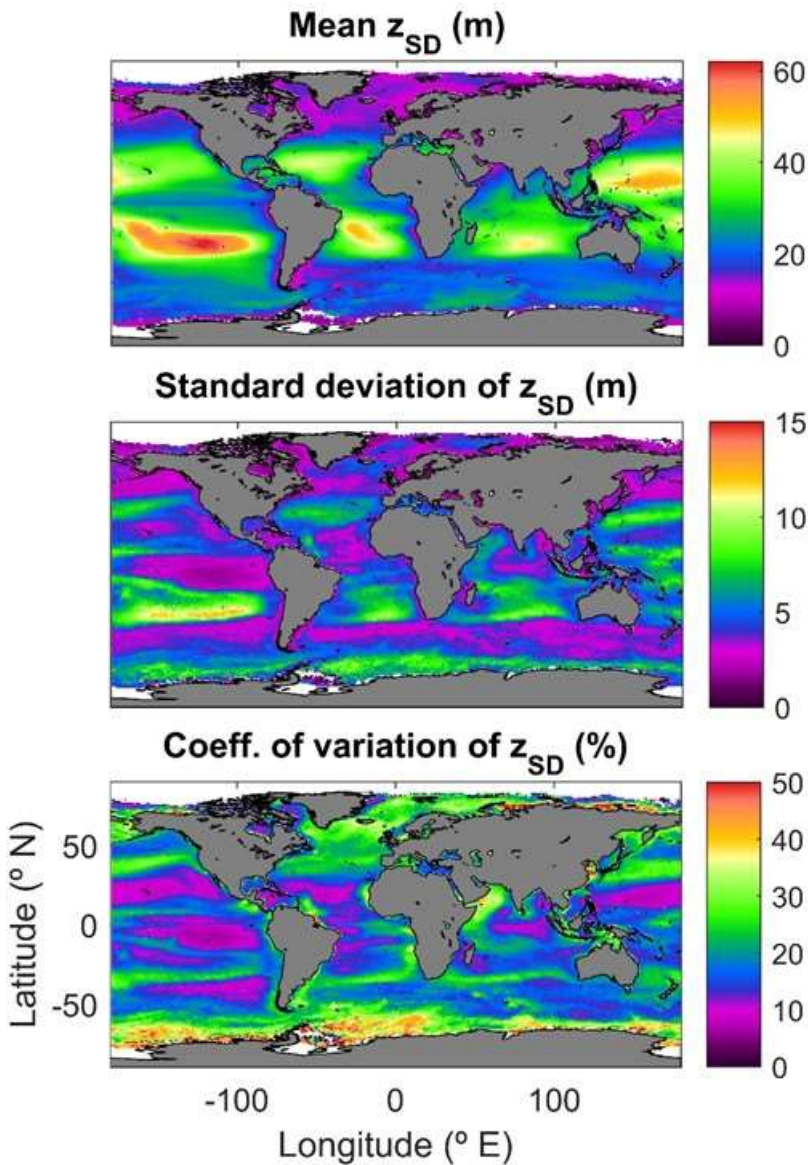


Figure 2: Global maps of the Secchi disk depth (z_{SD}) in the period 1997 - 2018. Top: series mean. Mid: standard deviation. Bottom: coefficient of variation.

260 The hue angle (α) shows a pattern similar to z_{SD} (Fig. 3), though with a marked non-linearity. Global values cover
 | the range $\sim 50^\circ - 235^\circ$, though 85% of the world's oceans are blue with a hue angle above $\geq 200^\circ$. The standard
 | deviation is very low in the oligotrophic areas, reflecting the fact that the human eye has little sensitivity to
 | variations in ~~color~~colour when these happen at the lower spectral end of the visible range. On the other hand,
 265 shelf seas and coastal zones.

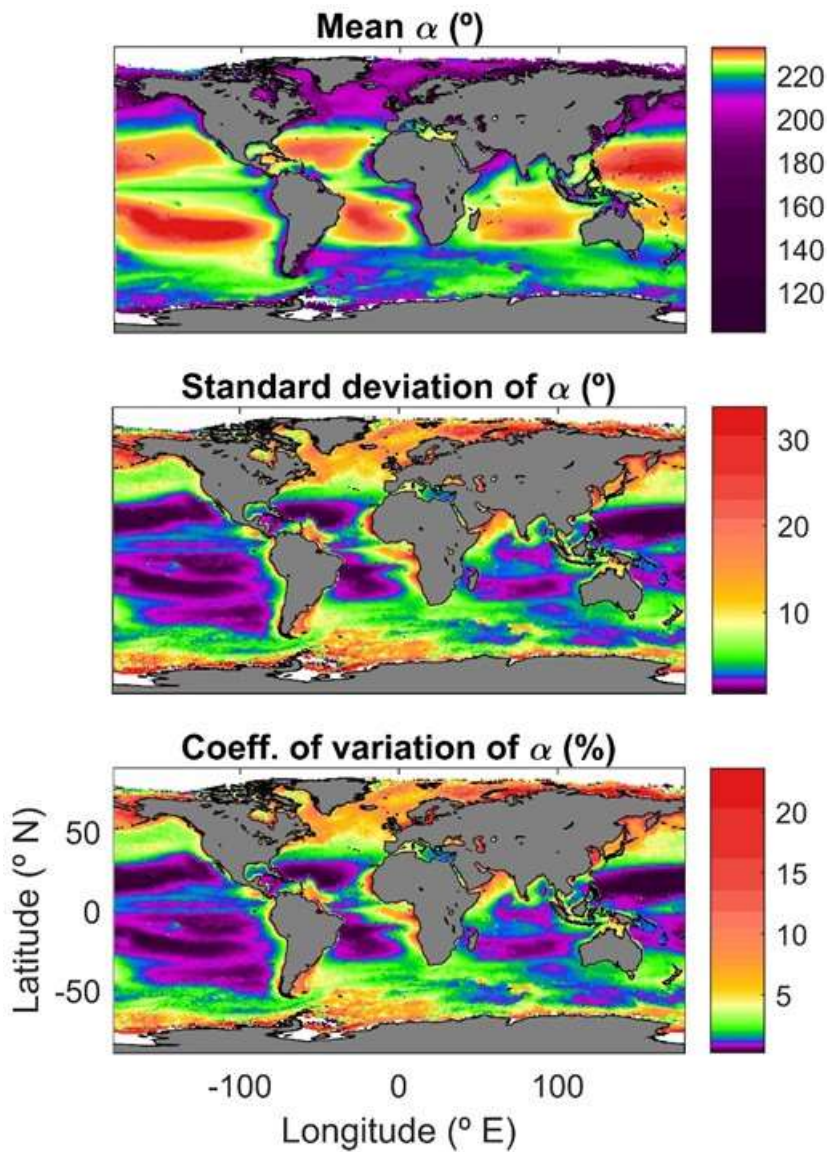


Figure 3: Global maps of the hue angle (α) in the period 1997 - 2018. Top: series mean. Mid: standard deviation. Bottom: coefficient of variation.

270 The Forel-Ule mean value (Fig. 4) reflects the known global patterns. Despite being-ranging the scale from 0 to 21, the-lowest 99 % of the data are within $FU \leq 8$, keeping the upper rest of the FU scale for very turbid coastal areas or highly humic waters. Note that, despite the fractional results due to statistics, FU are integers. As shown for the hue angle, FU shows very little variation for the oligotrophic and low latitude areas, reaching the widest dynamic range in coastal and high-latitude blooming seas.

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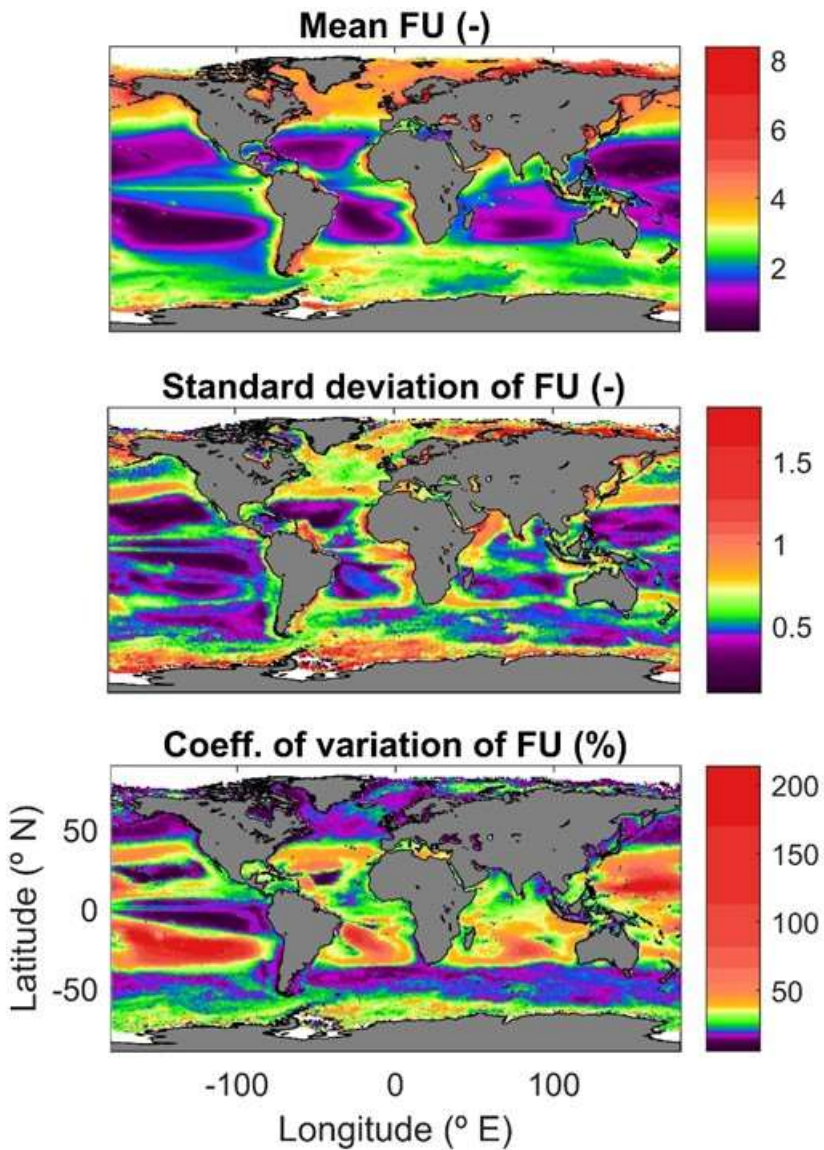
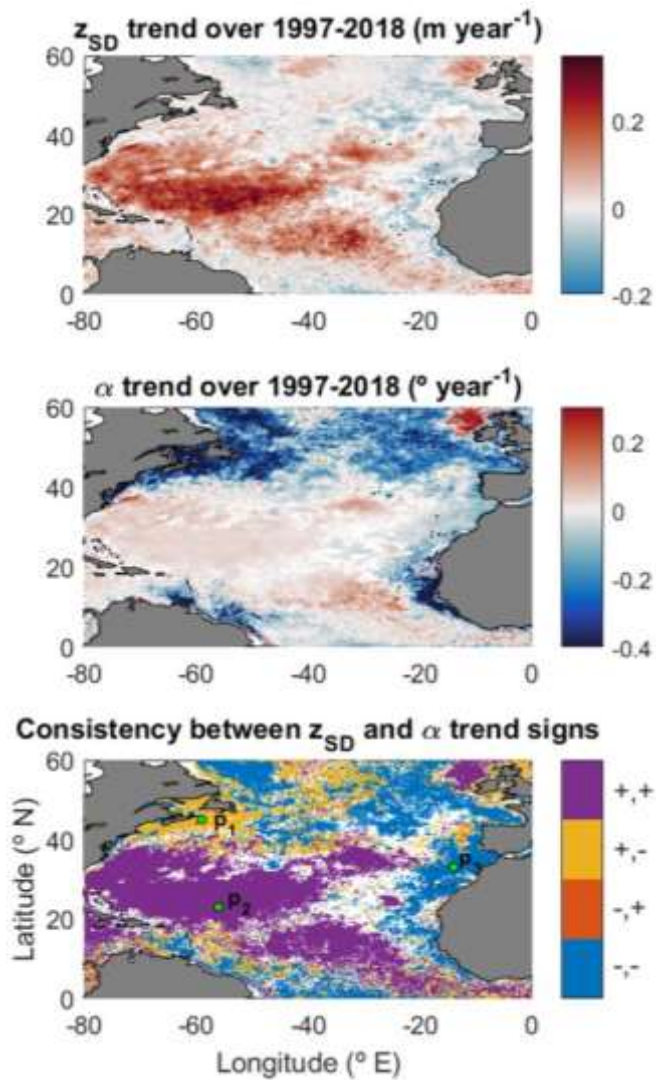


Figure 4: Global maps of the Forel-Ule (FU) index in the period 1997 - 2018. Top: series mean. Mid: standard deviation. Bottom: coefficient of variation.

280 6. Example of application: trend analysis for the North Atlantic Ocean

As already argued, a main goal of the OC-CCI project is the construction of a time series for the calculation of oceanic trends [and oscillations](#). Here, we used the lower spatial resolution product at 0.25° , covering the Atlantic Ocean north of the equator. This section reports [a](#) trend analysis of the area for the Secchi disk depth and the hue angle. Seasonal decomposition was made pixel-wise with the publicly available BEAST code (Zhao et al.,

285 2019). The Sen's slope (Sen, 1968) of the de-seasonalized signal was calculated and the result was tested for
 significance at 95 % (Mann, 1945; Kendall, 1975). In Fig. 5 the upper two panels only only show only the statistically
 significant trend signals for z_{SD} and α , calculated for the period 1997-2018. A positive (negative) trend in z_{SD} means
 that a Secchi disk will disappear from sight at a larger (shallower) depth, corresponding to a higher (lower)
 transparency increase (decrease). A positive (negative) trend in α indicates a hue angle means that the spectrum
 290 shifts towards the blue (green/red) shift, normally corresponding to a reduction (increase) in the surface
 chlorophyll-a concentration (Pitarch et al., 2019b). In the lowest panel of Fig. 5, the sign (positive or negative) of
 the trends are combined to yield 4 four classes; a combination of positive z_{SD} sign and negative hue angle sign are
 given the code (+ -). Again, no colour indicates no significant trend in both indexes.



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Figure 5: Upper figure: linear slope of Secchi disk (z_{SD}) trend at pixel scale for the North Atlantic. Central figure: same as before, but for the hue angle (α). Lower figure: signs of both linear slopes.

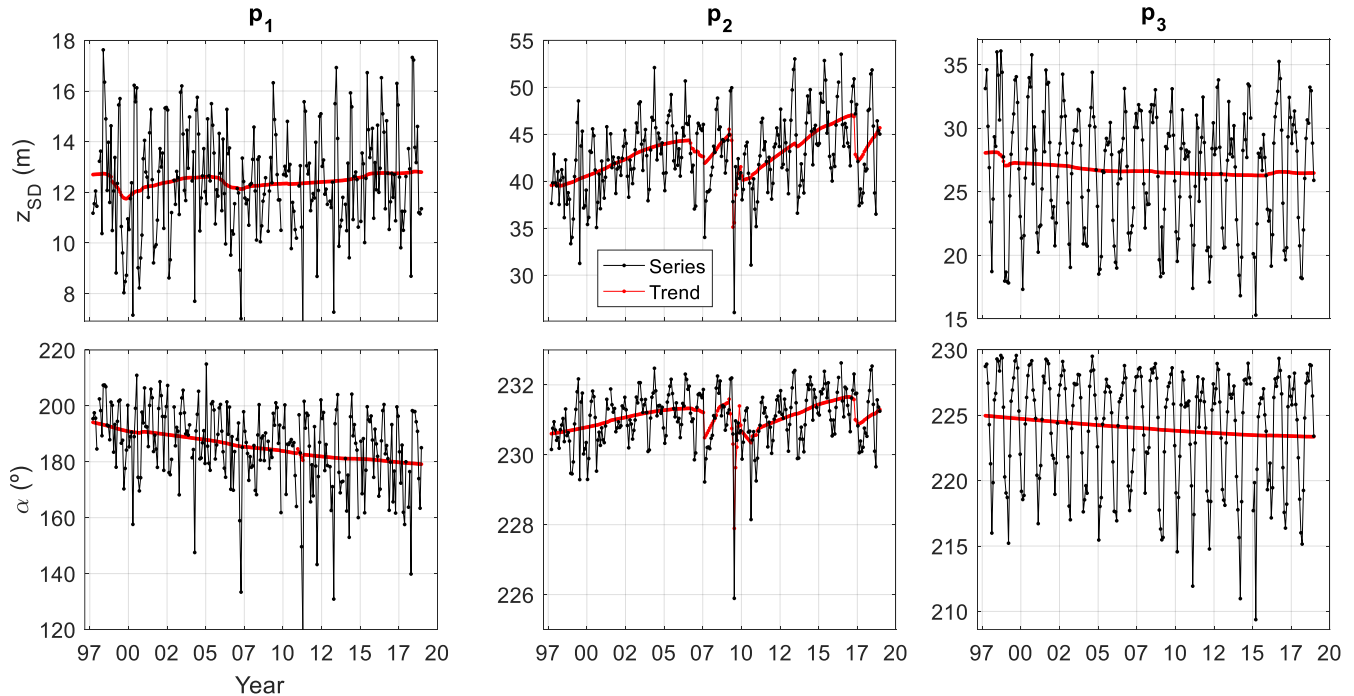


Figure 6: Time series of Secchi disk depth (upper panels) and hue angle (lower panels) at three points in the North Atlantic Ocean. Red lines show the ~~trend after de-seasonalization~~ seasonalized signal.

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The North-Atlantic trend maps (Fig. 5) show that the subtropical gyre is gaining transparency and ~~becoming getting bluer bluer and bluer~~. On the other hand, mesotrophic, more northern waters outside the gyre as well as coastal ~~w~~waters are experiencing totally different dynamics. In general, there is a loss of transparency and a ~~green red~~ shift. Some areas where trends are very small have apparent opposite trends in ~~z_{SD} and α~~ the Secchi disk depth and hue angle, which highlights the fact that ~~the Secchi disk depth and these two quantities hue angle~~

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do not completely contain the same information, and it will deserve some future research. In Fig. 6, the complexity in the spatial and temporal ~~behaviour~~ behavior is detailed for three points, their position indicated on the lower map of Fig 5. We anticipate that the study of these variations at various spatial scales and their relation to potential physical and environmental drivers like sea surface temperature (SST), ocean heat content, nutrients, mixed layer depth, coastal currents and river outflow will be explored.

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7. Data availability

The 256-full resolution netCDF files constituting the global time series at monthly frequency and 4 km resolution is publicly available through the following link: <https://doi.pangaea.de/10.1594/PANGAEA.904266> (Pitarch et al., 2019a). ~~This~~ The repository also contains two ~~teaser~~ sub-products of much smaller size that can be quickly

315 downloaded and easily handled locally: (1) a downgraded global dataset at 1-° for latitude and longitude and (2)
a downgraded dataset at 0.25-° for latitude and longitude, for the North Atlantic Ocean.

8. Conclusions

320 We have derived a consistently calibrated 21-year-long dataset of monthly Secchi disk depth, hue angle and Forel-
Ule index over the global oceans, at ~ 4 km spatial resolution at the equator. The algorithms employed for this
release have been recently published, are at the state of the art and their application is not restricted to any
specific water type. Best-effort validation statistics have been provided based on current evidence. Global
average maps have shown meaningful patterns that are consistent with the general understanding of the ocean
dynamics. A small example on trend estimation has also been provided to illustrate one of the main areas of
research that can be addressed with this dataset.

325 This dataset has been directly derived from the OC-CCI dataset, and as such, shares its advantages and limitations.
Main advantages are the careful sensor inter-calibration to avoid spurious trends, the constant revision of the
methodology in search for improved quality and the warranted continuation of the time series for the next years.
Limitations are the exclusion of high latitude areas due to high solar zenith angle, that causes problems in the OC-
CCI processing and the uneven monthly binning depending on cloud cover. These issues could be solved in a
330 future release, which could ~~start from~~ use daily source data with an application of a gap-filling procedure.

With this release, we hope to eliminate redundant efforts within the community to estimate water transparency
and colour from satellite measurements. Depending on the reception and user needs, future releases using other
temporal and spatial resolutions could be made.

335 ~~For z_{SD} -The Secchi disk depth and FUthe Forel-Ule index, these~~ time series could be further enlarged backwards
by including-linking them to the archived in-situ data that go back to the year 1890 (Boyce et al., 2012). However,
even in the absence of historic data, twenty-one years of satellite data are already a valuable dataset that may
provide new insights on the ocean's variability in relation to climate change and constitute valuable information
to train and validate mechanistic oceanic models.

9. Author contribution

340 J.P. conceived the manuscript, produced the dataset and associated results and wrote the first draft of the
manuscript, following advice by M.B. and S.M. M.B., H.vdW. and S.M. revisions of the draft. J.P. wrote the final
submitted version.

10. Competing interests

The authors declare that they have no conflict of interest.

345 11. Acknowledgements

J.P. thanks financial support by the Ministry for Science and Culture of Lower Saxony, Germany in the framework of the 'Coastal Ocean Darkening' project (VWZN3175) and by the Copernicus Climate Change Service (C3S_511). M. B. is supported by the ESA Living Planet Fellowship Project PHYSIOGLOB: Assessing the inter-annual physiological response of phytoplankton to global warming using long-term satellite observations, 2018-2020.

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References

- Boyce, D. G., Lewis, M., and Worm, B.: Integrating global chlorophyll data from 1890 to 2010, *Limnology and Oceanography: Methods*, 10, 840-852, <https://doi.org/10.4319/lom.2012.10.840>, 2012.
- 355 Ceccaroni, L., Piera, J., Wernand, M. R., Zielinski, O., Busch, J. A., Woerd, H. J. v. d., Bardaji, R., Friedrichs, A., Novoa, S., Thijsse, P., Velickovski, F., Blaas, M., and Dubsky, K.: Citclops: A next-generation sensor system for the monitoring of natural waters and a citizens' observatory for the assessment of ecosystems' status, *PLOS ONE*, 15, e0230084, <https://doi.org/10.1371/journal.pone.0230084>, 2020.
- IOCCG: Ocean-colour data merging, IOCCG, Dartmouth, Canada, No. 6, 2007.
- 360 Jackson, T., Chuprin, A., Sathyendranath, S., Grant, M., Zühlke, M., Dingle, J., Storm, T., Boettcher, M., and Fomferra, N.: Ocean Colour Climate Change Initiative (OC_CCI) –Interim Phase. Product User Guide. D3.4 PUG, 2019.
- Kendall, M.: Rank correlation methods. 2nd impression, Charles Griffin and Company Ltd. London and High Wycombe, 1975.
- Lee, Z., Carder, K. L., and Arnone, R. A.: Deriving inherent optical properties from water color: a multiband quasi-analytical algorithm for optically deep waters, *Appl. Opt.*, 41, 5755-5772, <https://doi.org/10.1364/AO.41.005755>, 2002.
- 365 Lee, Z., Hu, C., Shang, S., Du, K., Lewis, M., Arnone, R., and Brewin, R.: Penetration of UV-visible solar radiation in the global oceans: Insights from ocean color remote sensing, *Journal of Geophysical Research: Oceans*, 118, 4241-4255, <https://doi.org/10.1002/jgrc.20308>, 2013.
- Lee, Z., Shang, S., Hu, C., Du, K., Weidemann, A., Hou, W., Lin, J., and Lin, G.: Secchi disk depth: A new theory and mechanistic model for underwater visibility, *Remote Sensing of Environment*, 169, 139-149, <https://doi.org/10.1016/j.rse.2015.08.002>, 2015.
- 370 Mann, H. B.: Nonparametric Tests Against Trend, *Econometrica*, 13, 245-259, <https://doi.org/10.2307/1907187>, 1945.
- Nie, Y., Guo, J., Sun, B., and Lv, X.: An evaluation of apparent color of seawater based on the in-situ and satellite-derived Forel-Ule color scale, *Estuarine, Coastal and Shelf Science*, 246, 107032, <https://doi.org/10.1016/j.ecss.2020.107032>, 2020.
- Secchi disk and water color observations: <https://www.nodc.noaa.gov/OC5/WOD/secchi-data-format.html>, access: Aug. 11, 2013.
- 375 Novoa, S., Wernand, M. R., and van der Woerd, H. J.: The Forel-Ule scale revisited spectrally: preparation protocol, transmission measurements and chromaticity, *Journal of the European Optical Society - Rapid publications*; Vol 8 (2013), <https://doi.org/10.2971/jeos.2013.13057>, 2013.

- Pitarch, J.: Biases in ocean color over a Secchi disk, *Opt. Express*, 25, A1124-A1131, <https://doi.org/10.1364/OE.25.0A1124>, 2017.
- 380 Pitarch, J., Van Der Woerd, H. J., Brewin, R. J. W., and Zielinski, O.: Twenty years of monthly global maps of Hue angle, Forel-Ule and Secchi disk depth, based on ESA-OC-CCI data, PANGAEA, <https://doi.org/10.1594/PANGAEA.904266>, 2019a.
- Pitarch, J., Van Der Woerd, H. J., Brewin, R. J. W., and Zielinski, O.: Optical properties of Forel-Ule water types deduced from 15 years of global satellite ocean color observations, *Remote Sensing of Environment*, 231, 111249, <https://doi.org/10.1016/j.rse.2019.111249>, 2019b.
- 385 <https://doi.org/10.1016/j.rse.2019.111249>, 2019b.
- Pitarch, J.: A review of Secchi's contribution to marine optics and the foundation of Secchi disk science, *Oceanography*, 33(3), <https://doi.org/10.5670/oceanog.2020.301>, 2020.
- Pitarch, J., Bellacicco, M., Organelli, E., Volpe, G., Colella, S., Vellucci, V., and Marullo, S.: Retrieval of Particulate Backscattering Using Field and Satellite Radiometry: Assessment of the QAA Algorithm, *Remote Sensing*, 12, 77, <https://doi.org/10.3390/rs12010077>, 2020.
- 390 <https://doi.org/10.3390/rs12010077>, 2020.
- Polovina, J. J., Howell, E. A., and Abecassis, M.: Ocean's least productive waters are expanding, *Geophysical Research Letters*, 35, <https://doi.org/10.1029/2007g1031745>, 2008.
- Sen, P. K.: Estimates of the Regression Coefficient Based on Kendall's Tau, *Journal of the American Statistical Association*, 63, 1379-1389, <https://doi.org/10.1080/01621459.1968.10480934>, 1968.
- 395 Valente, A., Sathyendranath, S., Brotas, V., Groom, S., Grant, M., Taberner, M., Antoine, D., Arnone, R., Balch, W. M., Barker, K., Barlow, R., Bélanger, S., Berthon, J. F., Beşiktepe, Ş., Borsheim, Y., Bracher, A., Brando, V., Canuti, E., Chavez, F., Cianca, A., Claustre, H., Clementson, L., Crout, R., Frouin, R., García-Soto, C., Gibb, S. W., Gould, R., Hooker, S. B., Kahru, M., Kampel, M., Klein, H., Kratzer, S., Kudela, R., Ledesma, J., Loisel, H., Matrai, P., McKee, D., Mitchell, B. G., Moisan, T., Muller-Karger, F., O'Dowd, L., Ondrusek, M., Platt, T., Poulton, A. J., Repecaud, M., Schroeder, T., Smyth, T.,
- 400 Smyth-Wright, D., Sosik, H. M., Twardowski, M., Vellucci, V., Voss, K., Werdell, J., Wernand, M., Wright, S., and Zibordi, G.: A compilation of global bio-optical in situ data for ocean-colour satellite applications – version two, *Earth Syst. Sci. Data*, 11, 1037-1068, <https://doi.org/10.5194/essd-11-1037-2019>, 2019.
- Van Der Woerd, H. J., and Wernand, M. R.: True Colour Classification of Natural Waters with Medium-Spectral Resolution Satellites: SeaWiFS, MODIS, MERIS and OLCI, *Sensors*, 15, 25663-25680, <https://doi.org/10.3390/s151025663>, 2015.
- 405 Van Der Woerd, H. J., and Wernand, M. R.: Hue-Angle Product for Low to Medium Spatial Resolution Optical Satellite Sensors, *Remote Sensing*, 10, 180, <https://doi.org/10.3390/rs10020180>, 2018.
- Wang, S., Li, J., Zhang, B., Spyrakos, E., Tyler, A. N., Shen, Q., Zhang, F., Kuster, T., Lehmann, M. K., Wu, Y., and Peng, D.: Trophic state assessment of global inland waters using a MODIS-derived Forel-Ule index, *Remote Sensing of Environment*, 217, 444-460, <https://doi.org/10.1016/j.rse.2018.08.026>, 2018.
- 410 Wernand, M. R.: On the history of the Secchi disc, *Journal of the European Optical Society - Rapid publications*; Vol 5 (2010), <https://doi.org/10.2971/jeos.2010.10013s>, 2010.
- Wernand, M. R., and Van Der Woerd, H. J.: Spectral analysis of the Forel-Ule Ocean colour comparator scale, *Journal of the European Optical Society - Rapid publications*; Vol 5 (2010), <https://doi.org/10.2971/jeos.2010.10014s>, 2010.

415 Wernand, M. R., Hommersom, A., and Van Der Woerd, H. J.: MERIS-based ocean colour classification with the discrete
Forel–Ule scale, *Ocean Sci.*, 9, 477-487, <https://doi.org/10.5194/os-9-477-2013>, 2013.

Zhao, K., Wulder, M. A., Hu, T., Bright, R., Wu, Q., Qin, H., Li, Y., Toman, E., Mallick, B., Zhang, X., and Brown, M.:
Detecting change-point, trend, and seasonality in satellite time series data to track abrupt changes and nonlinear dynamics: A
Bayesian ensemble algorithm, *Remote Sensing of Environment*, 232, 111181, <https://doi.org/10.1016/j.rse.2019.04.034>, 2019.

420 Zibordi, G., Voss, K. J., Johnson, B. C., and Mueller, J. L.: *In Situ Optical Radiometry*, Dartmouth, NS, Canada, 2019.