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ABSTRACT

31 System Vicarious Calibration (SVC) ensures a relative radiometric calibration to satellite ocean color 32 sensors that minimizes uncertainties in the water-leaving radiance L_w derived from the top of atmosphere 33 radiance L_T. This is achieved through the application of gain-factors, g-factors, to pre-launch absolute 34 radiometric calibration coefficients of the satellite sensor corrected for temporal changes in radiometric 35 sensitivity. The g-factors are determined by the ratio of simulated to measured spectral $L_{\rm T}$ values where 36 the former are computed using: *i*. highly accurate *in situ L*w reference measurements; and *ii*. the same 37 atmospheric models and algorithms applied for the atmospheric correction of satellite data. By analyzing 38 basic relations between relative uncertainties of L_w and L_T , and g-factors consistently determined for the 39 same satellite mission using different *in situ* data sources, this work suggests that the creation of ocean 40 color Climate Data Records (CDRs) should ideally rely on: i. one main long-term in situ calibration 41 system (site and radiometry) established and sustained with the objective to maximize accuracy and 42 precision over time of g-factors and thus minimize possible biases among satellite data products from 43 different missions; and additionally *ii*. unique (i.e., standardized) atmospheric model and algorithms for 44 atmospheric correction to maximize cross-mission consistency of data products at locations different 45 from that supporting SVC. Finally, accounting for results from the study and elements already provided 46 in literature, requirements and recommendations for SVC sites and field radiometric measurements are 47 streamlined.

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Keywords: Ocean Color, System Vicarious Calibration, Climate Data Record

1. Introduction

In recent decades, measurements of ocean color from earth-orbiting satellite sensors have 51 52 demonstrated high value for a number of applications ranging from regional water quality assessment 53 (e.g., Attila et al. 2013) to global climate change investigations (e.g., Behrenfeld et al. 2006). Confidence 54 in results from these applications, however, depends on accuracy of the satellite-derived data products. 55 The primary ocean color product is the spectral water-leaving radiance $L_{\rm w}$, i.e., the radiance emerging 56 from the sea that is retrieved from the total radiance $L_{\rm T}$ detected by the satellite, whose accuracy 57 determines those of satellite-derived data products. These include the spectral distribution of the 58 normalized water-leaving radiance L_{WN} (i.e., the water-leaving radiance that would occur with no 59 atmosphere, the sun at the zenith and at the mean sun-earth distance) or of the equivalent remote sensing 60 reflectance R_{RS} , applied to determine geophysical quantities such as the near-surface chlorophyll-a 61 concentration (Chla).

Early accuracy requirements for satellite ocean color data products generally refer to the work of Gordon and Clark (1981), Gordon et al. (1983) and Gordon (1987). By considering oligotrophic waters, they indicated a 5% uncertainty for L_w in the blue spectral region to allow for the determination of *Chla* concentration with a 35% maximum uncertainty. Subsequently, spectrally independent uncertainties of 5%, with a 1% inter-band uncertainty, were included among the objectives of the Sea-viewing Wide Field-of-view Sensor (SeaWiFS) mission (Hooker et al. 1992). These target uncertainties were later retained for successive missions and have become a science requirement for the ocean color community.

Achievement of the spectrally independent 5% uncertainty target in satellite-derived L_w is mostly challenged by the accuracy of the absolute radiometric calibration of satellite optical sensors and additionally by uncertainties in the quantification of the large atmospheric perturbations affecting L_T . In particular, current uncertainties of approximately 2-3% (Butler et al. 2007, Eplee et al. 2011, Esposito et al. 2004) in the absolute radiometric calibration of satellite sensors in the visible spectral region and the additional uncertainties affecting the atmospheric correction process generally larger than a few percent (IOCCG 2010), may lead to large differences among multi-mission L_w data (Zibordi et al. 2006). 76 These limitations can be resolved through the so-called System Vicarious Calibration (SVC) that 77 determines vicarious adjustment gain-factors g (hereafter g-factors) for absolute radiometric calibration 78 coefficients of satellite sensors (Gordon 1998) through simulation of top-of-atmosphere $L_{\rm T}$ using: *i*. 79 highly accurate in situ Lw reference measurements; and ii. the same atmospheric models and algorithms 80 as applied for the atmospheric correction of satellite data. The g-factors, determined by the ratio of 81 simulated to measured spectral $L_{\rm T}$ values, aim at minimizing the combined effects of: *i*. uncertainties due 82 to the absolute pre-flight radiometric calibration and characterization of the satellite sensor after correction for sensitivity change with time; and *ii*. inaccuracy of the models and algorithms applied in 83 the atmospheric correction to determine L_w from L_T . Clearly, the SVC process allows the determination 84 of L_w with the lowest uncertainty when satellite observation conditions are equivalent to those 85 86 characterizing the data applied for the calculation of g-factors (i.e., when mean biases affecting the SVC 87 and the regular atmospheric correction processes are identical and fully compensate each other). It must 88 be emphasized that the system nature of SVC requires re-computing g-factors after any change in the 89 models or algorithms applied for the atmospheric correction, or any significant change in instrument 90 absolute or temporal calibration knowledge.

By considering uncertainty requirements for satellite-derived L_w applicable for the construction of Climate Data Records (CDRs), which serve as core climate benchmark observations, the present work investigates the calibration needs for L_T with the objective of discussing requirements for *in situ* L_w data suitable for the determination of *g*-factors.

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2. System Vicarious Calibration Requirements

98 Vicarious calibration broadly refers to the indirect calibration of satellite sensors through simulation of 99 top-of-atmosphere data (Koepke 1982). Generic vicarious calibration methods based on atmospheric 100 models and algorithms different from those applied for the operational data processing cannot reduce 101 absolute uncertainties in derived radiometric calibration factors below a few percent (IOCCG, 2013). 102 This may lead to very large uncertainties in satellite-derived L_w (see §2.1). Consequently, unlike SVC 103 that minimizes uncertainties in retrieved L_w (Gordon 1998), generic vicarious calibration methods are 104 best applied for the quality check of pre-launch absolute radiometric calibrations of satellite ocean color 105 sensors.

106 In view of supporting the discussion on accuracy and precision needs for *g*-factors from SVC, the 107 following subsections will review: *i*. requirements for the construction of CDRs from satellite-derived 108 L_w ; and *ii*. legacy requirements for *in situ* L_w reference measurements.

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110 2.1 Requirements for CDRs of L_w

CDRs of Essential Climate Variables (ECVs) are intended to support climate change investigations
 through time-series of core benchmark observations with enough accuracy to allow the detection of long-

term trends embedded in large natural variations (Leroy et al. 2008).

114 Requirements for the generation of a CDR of satellite-derived L_{WN} from L_w (WMO 2011), which is the 115 fundamental satellite ocean color ECV, include:

Radiometric uncertainty lower than 5% in the blue and green spectral regions (downscaled with
 respect to the spectrally independent 5% uncertainty target listed among the objectives of several ocean
 color missions);

119 2. Stability better than 0.5% over a decade.

The requirement on uncertainty is essential to understand climate-driven processes and changes, while
the requirement on stability is essential to confidently determine long-term changes or trends (Ohring et
al. 2005).

As already anticipated, the strict requirement of 5% maximum uncertainty for L_w determined from L_T at relevant wavelengths, requires the application of SVC. While this need is commonly accepted by the satellite ocean color community, the accuracy and precision required for *g*-factors for different missions supporting the creation of CDRs appears less consolidated. To strictly address such a need, the relationship linking uncertainties in L_w and L_T is hereafter investigated through the use of the measurement equation. Specifically, in the absence of atmospheric gaseous absorption and sun glint and foam perturbations, the top-of-atmosphere radiance L_T can be related to L_w through the following simplified model

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$$L_{\rm T} = L_{\rm R} + L_{\rm A} + L_{\rm w} t_{\rm d} \tag{1}$$

where $L_{\rm R}$ and $L_{\rm A}$ indicate the Rayleigh and aerosol atmospheric radiance contributions, and $t_{\rm d}$ is the diffuse atmospheric transmittance that varies with atmospheric path-length and constituents. By assuming the values of $L_{\rm R}$ and $L_{\rm A}$ are exactly determined for any given observation condition, following the Guide to the Expression of Uncertainty in Measurement (JCGM, 2008) Zibordi and Voss (2014) provided equations relating absolute uncertainties of $L_{\rm T}$, $u(L_{\rm T})$, to those of $L_{\rm w}$, $u(L_{\rm w})$, and also linking relative uncertainties $u(L_{\rm T})/L_{\rm T}$ to $u(L_{\rm w})/L_{\rm w}$. In agreement with their work, $u(L_{\rm T})$ and $u(L_{\rm T})/L_{\rm T}$ are given by 139

$$u(L_T) = u(L_w)t_d \tag{2}$$

141 and

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$$\frac{u(L_T)}{L_T} = \frac{u(L_W)}{L_W} t_d \frac{L_W}{L_T}.$$
 (3)

143 For the purpose of this study centered on SVC, the uncertainties related to the atmospheric correction 144 process do not influence the determination of L_w because both SVC and the atmospheric correction rely 145 on the same robust atmospheric models and algorithms. Thus, to a first approximation Eq. 2 indicates 146 that the absolute uncertainties $u(L_T)$ and $u(L_w)$ are solely related by the factor t_d . Differently, Eq. 3 shows 147 that relative uncertainties $u(L_T)/L_T$ and $u(L_w)/L_w$ are additionally related by the ratio L_w/L_T . Because of 148 this, while the relation between absolute uncertainties only slightly varies with the atmospheric optical 149 properties through t_d , the dependence between relative uncertainties is highly variable with both marine 150 and atmospheric optical properties, which affect the term $t_d L_w/L_T$. Thus, while satellite-derived L_w may

exhibit similar absolute uncertainties for data collected over different water types, the corresponding relative uncertainties may largely differ as a function of L_w and L_T . Considering that requirements for satellite ocean color CDRs are provided in relative terms (e.g., see Ohring et al. 2005 and WMO 2011), the following analysis only focuses on relative uncertainties.

Rearranging Eq. 3 as a function of $u(L_T)/L_T$, for which a realistic spectrally independent radiometric uncertainty of 2% is assumed together with an ideal value of $t_d=1$, $u(L_w)/L_w$ would be approximately 20%, 40% and 200% for L_w/L_T equal to 0.10, 0.05 and 0.01, respectively. These uncertainty values, which may tentatively refer to blue, green and red wavelengths in oligotrophic waters, show the impossibility of meeting science requirements when only relying on current absolute radiometric calibration uncertainties, even assuming an exact quantification of the atmospheric perturbations.

161 Conversely, the application of Eq. 3 assuming $t_d=1$ and a spectrally independent uncertainty of 5% for 162 $L_{\rm w}$, implies values of $u(L_{\rm T})/L_{\rm T}$ as low as 0.5%, 0.25% and 0.05% for $L_{\rm w}/L_{\rm T}$ equal to 0.10, 0.05 and 0.01, 163 respectively. These values provide an estimate for the required spectral uncertainties of absolute radiometric calibrations for satellite ocean color sensors and further confirm that: *i*. even assuming that 164 the uncertainties in $u(L_w)/L_w$ due to atmospheric correction are negligible, the sole uncertainties currently 165 affecting in-flight absolute radiometric calibration are an impediment to meet ocean color science 166 167 requirements for CDRs; and that *ii*. SVC is the only viable alternative to overcome limitations due to uncertainties in absolute radiometric calibration and atmospheric correction. It is additionally observed 168 169 that, even accounting for future developments in absolute radiometric calibration, that are expected to considerably reduce uncertainties (Cramer et al. 2013 ans Levick et al. 2014), SVC will still remain an 170 essential component of any ocean color mission to minimize effects of inaccurate atmospheric 171 172 corrections.

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174 2.2 Legacy Requirements for SVC sites and data

Early indications on the appropriateness of SVC sites for global missions (mostly derived from Gordon(1998)) include:

i. Cloud free, very clear and maritime atmosphere with aerosol optical thickness $\tau_a < 0.1$ in the visible, which maximizes the potential number of satellite and *in situ* coincident data (i.e., matchups) and additionally optimizes the performance of the atmospheric correction process;

180 ii. Horizontally uniform L_w over spatial scales of a few kilometers to increase the comparability 181 between satellite and *in situ* data at different spatial resolutions;

182 iii. Mesotrophic waters to minimize the effects of *in situ* L_w measurement uncertainties in the blue 183 spectral region (this requirement has been considered less stringent with respect to the previous two, 184 leding to consider oligotrophic waters as an appropriate alternative);

iv. Coincident aerosol measurements to assess the atmospheric correction process.

186 In situ L_w data applicable for SVC are expected to have low uncertainty through the application of 187 state-of-the-art instrumentation, data reduction and quality assurance/control. Indications, mostly derived 188 from Clark et al. (2002), include the need for:

i. Hyper-spectral measurements to cover any ocean color spectral band regardless of its center-wavelengths and spectral responses;

ii. Fully characterized *in situ* radiometers to minimize uncertainties and allow their comprehensivequantification;

193 iii. Traceability of data to the International System of Units (SI) to ensure consistency with community194 shared measurement methods and standards.

Also, in the case of global data products contributing to the construction of CDRs, SVC should be applied using *in situ* L_w from measurement sites representative of the most common satellite observation conditions, i.e., the world oceans. The determination of regional *g*-factors has also been proposed for areas exhibiting unique optical features (Franz et al. 2007). It is, however, recognized that this solution is mostly intended to support local applications where accurate *in situ* L_w data exist.

Ultimately, the limited number of highly accurate *in situ* data and their high costs challenge SVC at large. This has generated debates on the suitability of a number of data sources for SVC and also motivation for various studies to explore legacy requirements. These studies have produced a number of 203 *g*-factors for the same satellite sensor relying on equivalent versions of the atmospheric correction code, 204 but using L_w from different sources. As will be shown later, results offer the great opportunity to 205 investigate differences among actual *g*-factors in view of discussing implications for the creation of 206 CDRs.

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209 **3.** Literature data

Among *in situ* systems specifically designed to support SVC for satellite ocean color sensors, only two 210 211 ensured almost continuous data collection across a number of satellite ocean color missions. These are: i. the Marine Optical Buoy (MOBY) developed by the National Oceanic and Atmospheric Administration 212 213 (NOAA) and the National Aeronautics and Space Administration (NASA) for the SeaWiFS and the 214 Moderate Resolution Imaging Spectroradiometer (MODIS) (Clark et al. 1997); and *ii*. the Buoy for the 215 Acquisition of a Long-Term Optical Time Series (Bouée pour L'acquisition de Séries Optiques à Long 216 Terme, BOUSSOLE), developed for the Medium Resolution Imaging Spectrometer (MERIS) by the Laboratoire d'Océanographie de Villefranche (LOV) in collaboration with a number of agencies 217 218 (Antoine at al. 2008a).

219 Aside from MOBY and BOUSSOLE (Eplee et al. 2001, Franz et al. 2007, Bailey et al. 2008), a number of alternative data sources were considered for SVC of SeaWiFS data (see Table 1). These 220 221 included in situ data sets obtained by combining measurements from a variety of instruments and 222 reduction schemes (Bailey et al. 2008), data from specific coastal areas commonly applied for regional 223 investigations (Mélin and Zibordi 2010), as well as modeled data (Werdell et al. 2007). Derived g-224 factors, consistently determined by applying the scheme detailed in Franz et al. (2007) and the SeaWiFS 225 Data Analysis System (SeaDAS, Fu et al. 1998) software package (version 5), are summarized in Table 226 2.

In agreement with Franz et al. (2007) and with specific reference to SeaWiFS center-wavelengths, *g*factors are assumed fixed and equal to unity at 865 nm, while the value at 765 nm is computed by

imposing a pure maritime aerosol model for locations in the oligotrophic gyres of the southern hemisphere. Spectral *g*-factors in the visible, which are those listed in Table 2, are successively determined from the average of individual factors computed imposing *in situ* reference water-leaving radiances as target values for the satellite-derived L_w . It is important to note that the averaging reduces the effects of random contributions to uncertainties in *g*-factors, but it does not remove the effects of any bias.

235 Recalling that unity g-factors indicate no correction, the values in Table 2 exhibit high consistency with differences generally within a few tenths of percent. The standard deviation, σ_g , gives an indication 236 237 of the precision affecting the SVC process as mostly resulting from *in situ* radiometer stability or varying 238 observation conditions. It is noted that the number of matchups used for SVC in all cases is larger than 239 the approximate 40 estimated by Franz et al. (2007) to determine sufficiently precise g-factors for 240 SeaWiFS using MOBY data. However, it is expected that such a number, implicitly referred to 241 SeaWiFS-MOBY matchups, may change when considering observation conditions different from those 242 offered by the MOBY site or satellite sensor performances different from those of SeaWiFS.

General elements on the various data sources utilized for the determination of the *g*-factors listed inTable 2 are summarized in the following sub-sections.

- 245
- Table 1. General elelments on the various sources utilized for SVC of SeaWiFS data: measurement
- 247 method, spectral features and site location (see text for additional details).

Data Source	L _w Method	Spectral Features	Site
MOBY	In-water, fixed depths	Hyper-spectral	Pacific Ocean (Hawaii)
MOBY-MS	In-water, fixed depths	Reduced resolution	Pacific Ocean (Hawaii)
BOUSSOLE	In-water, fixed depths	Multi-spectral	Ligurian Sea
NOMAD	Various	Various	Various
AAOT	Above-water	Multi spectral	Adriatic Sea
HOT-ORM	Modeled	User definable	Pacific Ocean (Hawaii)
BATS-ORM	Modeled	User Definable	Atlantic Ocean (Bermuda)

Table 2. Values of *g*-factors (*g*) and related standard deviations (σ_g) determined for SeaWiFS at its center-wavelengths. *N* indicates the number of matchups used for their determination, and *Y* the approximate number of measurement years.

	Wav	elength	41	2	44	3	49	0	51	.0	55	55	67	' 0
Data Source	Y	N	g	σ_{g}	g	σ_{g}	g	σ_{g}	g	σ_{g}	g	$\sigma_{\rm g}$	g	$\sigma_{\rm g}$
⁽¹⁾ MOBY	7	166	1.0368	0.009	1.0132	0.009	0.9918	0.008	0.9982	0.009	0.9993	0.009	0.9729	0.007
(1) MOBY-MS	7	166	1.0401	0.009	1.0136	0.009	0.9949	0.008	0.9937	0.009	0.9958	0.009	0.9691	0.007
(1) BOUSSOLE	3	46 ⁽⁴⁾	1.0402	0.005	1.0129	0.027	0.9961	0.033	1.0015	0.031	1.0007	0.021	0.9672	0.006
⁽¹⁾ NOMAD	7	64	1.0395	0.013	1.0135	0.013	0.9967	0.014	0.9962	0.017	0.9989	0.013	0.9693	0.009
(2) AAOT	5	99	1.0425	0.012	1.0143	0.014	0.9969	0.018	0.9977	0.019	1.0034	0.022	0.9819	0.020
⁽³⁾ HOT-ORM	7	176	1.0300	0.015	1.0086	0.012	0.9879	0.009	0.9979	0.008	1.0046	0.009	0.9718	0.006
(3) BATS-ORM	7	241	1.0345	0.018	1.0020	0.016	0.9814	0.013	0.9941	0.011	1.0016	0.011	0.9731	0.006

252 ⁽¹⁾ Bailey et al. (2008), ⁽²⁾ Mélin and Zibordi (2010), ⁽³⁾ Werdell et al. (2007). ⁽⁴⁾ 5 matchups at 412 nm, only.

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254 3.1 MOBY and MOBY-MS

Since 1997, MOBY has been deployed approximately 11 nautical miles from Lanai (Hawaii) in 1200 m water depth (Clark et al. 1997, Clark et al. 2002). The site was selected based on requirements for an ideal SVC location and accounting for the need to ensure economical and convenient access to shore facilities.

The main components of the MOBY system are: *i*. a spar buoy tethered to a moored buoy; and *ii*. a 259 260 hyper-spectral radiometer operating in the 340-955 nm spectral region with 1 nm resolution, coupled via 261 fiber optics to a number of radiance and irradiance collectors. These collectors ensure measurements of in-water downward irradiance and upwelling radiance at 1, 5 and 9 m depth. Above-water downward 262 263 irradiance is additionally measured at 2.5 m above the sea surface. The MOBY radiometer system 264 undergoes regular characterizations and calibrations to guarantee high accuracy and traceability of data 265 to the US National Institute of Standards and Technology (NIST). Internal system sources allow daily monitoring of radiometric stability. By statistically combining uncertainty contributions including those 266 267 related to the calibration source and its transfer, radiometric stability during deployments, and

environmental effects, Brown et al. (2007) showed the capability of reducing uncertainties to 268 269 approximately 3% in the 412-666 nm spectral interval for upwelling radiance $L_{\rm u}$ used to determine $L_{\rm w}$. 270 A total of 166 MOBY-SeaWiFS matchups fulfilling strict SVC criteria (Bailey and Werdell 2006, 271 Franz et al. 2007) over a 7-year period, were applied by Bailey et al. (2008) to produce the SeaWiFS g-272 factors. Criteria for the inclusion of SeaWiFS data resulting from the average of L_w values from the 5×5 273 pixels centered at the MOBY site, are: no processing flag raised (e.g., indicating cloud contamination, 274 glint perturbations, navigation problems or failure of the atmospheric correction); satellite viewing angle less than 56 degrees; sun zenith angle less than 70 degrees; Chla lower than 0.2 μ g l⁻¹; τ_a in the near 275 infrared lower than 0.15; and coefficient of variation less than 0.15 for L_{WN} in the blue-green spectral 276 regions and τ_a in the near-infrared. It is anticipated that similar matchup selection criteria were applied to 277 278 the other datasets included in this review.

The qualified matchups were constructed by convolving MOBY hyperspectral L_w data with the actual SeaWiFS spectral band responses. Bailey et al. (2008) also considered the parallel case of MOBY L_w averaged over 10 nm bandwidths with center-wavelengths corresponding to those of SeaWiFS. These *g*factors, referred to as MOBY-Multispectral (MOBY-MS), provide the unique opportunity to look into changes only due to differences in spectral resolution. In fact the radiometer system and measurement conditions are exactly the same for both hyperspectral and derived multispectral data.

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286 *3.2 BOUSSOLE*

BOUSSOLE, operated in the Ligurian Sea since 2003, has been deployed at approximately 32 nautical miles from the coast in 2440 m water depth and relies on a moored buoy optimized to maximize its vertical stability and minimize the shading effects of its superstructure (Antoine et al. 2008a). Optical instrumentation on the buoy includes 7-band commercial radiometers with 10 nm bandwidth in the 400-700 nm spectral region. In–water upwelling radiance, upward irradiance, and downward irradiance are measured with radiometers deployed at 4 and 9 m depths, while the downward irradiance is also measured at 4 m above the sea surface. Spectrally independent uncertainty values of approximately 6% have been declared for the normalized remote sensing reflectance determined from L_w (Antoine et al. 2008b). Since 2008, BOUSSOLE is also equipped with hyperspectral radiometers to measure the inwater upwelling radiance and downward irradiance, and the above-water downward irradiance. Data from these instruments, which are not part of this study, will be relevant for vicarious calibration activities of future missions.

299 A significant difference characterizes the extrapolation methods applied to subsurface radiometric data from MOBY and BOUSSOLE. While MOBY values are simply determined from the linear fit of log-300 301 transformed radiometric measurements with respect to depth, BOUSSOLE sub-surface values result 302 from the propagation of the 4 and 9 m depth values to the surface through models. This latter data 303 reduction scheme, requiring estimates of Chla, takes into account Raman effects and the related 304 nonlinearity of the log-transformed radiometric measurements with depth. Differences between the linear 305 fits of log-transformed radiometric measurements and modelled values, are within a few percent at 412 306 nm but increase up to several tens percent at 670 nm (Antoine et al. 2008b).

BOUSSOLE data were also considered by Bailey et al. (2008) for the determination of SeaWiFS *g*factors. Specifically, 46 matchups were identified from approximately a 3-year data record by relaxing the inclusion criteria on *Chla* (0.25 instead of 0.20 μ g l⁻¹). However, only 5 matchups were available for the 412 nm center-wavelength due to unavailability of the spectral band during some deployments.

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312 *3.3 NOMAD*

The NASA bio-Optical Algorithm Data set (NOMAD, Werdell and Bailey 2005) includes multi-site and multi-source data resulting from the reprocessing and strict quality control of radiometric measurements from the SeaWiFS Bio-Optical Archive and Storage System (SeaBASS). The variety of measurement methods, instruments, calibration and also data reduction schemes, make it difficult to assign well-defined uncertainties to the NOMAD radiometric data set.

The SeaWiFS *g*-factors determined from NOMAD (Bailey et al. 2008) were computed using 64 matchups fulfilling SVC selection criteria –out of a total of 1039. These field radiometry data result from overall 3475 quality controlled measurements out of 15400 from 1350 field campaigns included in SeaBASS. These numbers clearly indicate the difficulty of supporting SVC with *in situ* L_w data from repositories constructed for applications more focused on validation and development rather than vicarious calibration.

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325 *3.4 AAOT*

326 In contrast with MOBY and BOUSSOLE data, which are collected with systems specifically designed to support SVC, time-series data from a number of globally distributed coastal sites established to 327 328 support satellite ocean color validation activities are accessible through the Ocean Color component of 329 the Aerosol Robotic Network (AERONET-OC, Zibordi et al. 2009). AERONET-OC field radiometers perform multispectral L_w measurements at a number of ocean color bands with center-wavelengths in the 330 331 410-1020 nm spectral region and 10 nm bandwidth. Data collection, reduction and quality control rely 332 on standardized methods (Zibordi et al. 2009) assuring cross-site consistency to data products. Among 333 AERONET-OC sites, the Acqua Alta Oceanographic Tower (AAOT, often indicated as 'Venise'), located in the northern Adriatic Sea at approximately 8 nautical miles from the main land, since 2003 has 334 provided an almost uninterrupted series of data largely applied for the validation of multi-mission ocean 335 336 color radiometric data (e.g., Zibordi et al. 2006, Mélin et al. 2011, Zibordi et al. 2012). Uncertainties of 337 5% in the blue-green and 8% in the red spectral regions have been quantified for the AAOT fully quality 338 assured normalized water-leaving radiance determined from $L_{\rm w}$ (Gergely and Zibordi 2014).

AERONET-OC data from the AAOT were used by Mélin and Zibordi (2010) for the determination of regional SeaWiFS *g*-factors. Specifically, 99 qualified matchups were identified from a 5-year data set by relaxing some selection criteria (e.g., accepting *Chla* up to 3 μ g l⁻¹ and coefficient of variation up to 0.20 in the blue-green spectral region for satellite data). A particular effort was devoted to correct *in situ L*_w spectra for the effects of differences in center-wavelengths with respect to SeaWiFS bands. Results from the study give insight on the relevance of coastal vicarious calibration sites for regional investigations and additionally provide elements to evaluate their suitability for global applications. Still, the spatial and inter-annual variability of both atmospheric and water optical properties in the region do not support the selection of the AAOT as a SVC site for the creation of CDRs.

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349 3.5 HOT-ORM and BATS-ORM

Ocean Reflectance Models (ORM) are an additional source of radiance spectra (Morel and Maritorena, 2001) expected to be of suitable accuracy for oceanic waters. Even though these models are mostly relevant for bio-optical investigations or as diagnostic tools, their usefulness for SVC has been investigated by Werdell et al. (2007) to verify their fitness for historical satellite ocean color sensors (i.e., CZCS and OCTS) for which an extensive time-series of *in situ* radiometric measurements do not exist.

The SeaWiFS *g*-factors determined using ORM methodology include those relying on the *Chla* timeseries from the U.S. Joint Global Ocean Flux Study (JGOFS) Bermuda Atlantic Time-series Study (BATS) and Hawaiian Ocean Time-series (HOT). Specifically, ORM-BATS *g*-factors were determined using 241 matchups from 1998 to 2004, while ORM-HOT *g*-factors were computed for the same period with 176 matchups (Werdell et al. 2007). Comprehensive uncertainty estimates for modeled L_w were not provided.

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363 4. Analysis and discussion

It shall be noted that the *g*-factors in Table 2 were determined with an earlier version of the SeaDAS processor (i.e., version 5) based on an atmospheric model and pre-launch absolute calibration factors (specifically at 412 nm) different from those currently in use. Because of this, the *g*-factors in Table 2 need to be considered outdated for present SeaWiFS data processing. Still, they are the result of a unique combination of investigations and remain a convenient data set to explore effects of differences among *g*-factors in the creation of CDRs. Making use of these data, the following analysis focuses on percent 370 differences between *g*-factors determined from MOBY data, g^{MOBY} , and those from other data sources, *g*, 371 computed as

$$\Delta g = 100 \ \frac{g - g^{MOBY}}{g^{MOBY}} \tag{4}$$

The choice of the *g*-factors from MOBY as the reference is justified by its ideal location (exhibiting oligotrophic waters and maritime aerosol, in addition to annual cycles of small amplitude) and an extensive characterization of field radiometers and careful examination of radiometric uncertainties. This choice, however, has not to be interpreted as implicitly advocating the use of MOBY for SVC of any satellite ocean color mission.

For completeness it is also mentioned that the HOT-ORM and BATS-ORM *g*-factors included in Table 2, were discussed by Werdell et al. (2007) with respect to the older MOBY *g*-factors determined by Franz et al. (2007) on the basis of 150 match-ups. Those *g*-factors exhibit spectrally averaged differences of -0.09% with respect to the more recent values by Bailey et al. (2008) used in the current analysis. Still, the changes in the values of Δg for HOT-ORM and BATS-ORM resulting from the application of the *g*-factors from Franz et al. (2007) instead of those from Bailey et al. (2008), does not affect the following discussion and conclusions.

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Table 3. Relative percent differences Δg between SeaWiFS *g*-factors determined using Eq. 4 applied to data in Table 2. The values in bold indicate Δg exceeding ±0.3% in the blue-green spectral regions and ±0.1% in the red.

Wavelength	412	443	490	510	555	670
Data Source	Δg [%]	Δg [%]	Δg [%]	Δg [%]	Δg [%]	Δg [%]
MOBY-MS	+0.32	+0.04	+0.31	-0.45	-0.35	-0.39
BOUSSOLE	+0.33	-0.03	+0.43	+0.33	+0.14	-0.59
NOMAD	+0.26	+0.03	+0.49	-0.20	-0.04	-0.37
AAOT	+0.55	+0.11	+0.51	-0.05	+0.41	+0.93
HOT-ORM	-0.66	-0.45	-0.39	-0.03	+0.53	-0.11
BATS-ORM	-0.22	-1.11	-1.05	-0.41	+0.23	+0.02

390 The Δg values in Table 3 from the same data source (i.e., inter-band) or across data sources (i.e., intra-391 band) are generally lower than $\pm 0.5\%$.

At a first scrutiny, the values of Δg determined for the AAOT and HOT-ORM appear to slightly differ from those determined for a more ideal site like BOUSSOLE or from a very large pool of data like NOMAD. Also interesting are the values of Δg determined for MOBY-MS, which clearly indicate the appreciable effects of non-matching spectral bands or SeaWiFS out-of-band responses, and consequently the importance of *in situ* hyperspectral L_w data.

Excluding HOT-ORM and BATS-ORM, the values of Δg exhibit high intra-band consistency between 412 and 490 nm, while they show a larger spread between 510 and 670 nm. Excluding a few spectral values from HOT-ORM (i.e., 412 nm), BATS-ORM (i.e., 443 and 490 nm) and AAOT (i.e., 670 nm), Δg is generally lower than ±0.5% for all the data sources.

401 In view of more quantitatively investigating differences in g-factors, Eq. 3 is applied to compute 402 $u(L_{\rm T})/L_{\rm T}$ as a function of $u(L_{\rm w})/L_{\rm w}$ accounting for actual mean spectral values of the term $t_{\rm d} L_{\rm w}/L_{\rm T}$ determined using 1997-2010 SeaWiFS data for three different locations: the MOBY site in the Pacific 403 Ocean with mean satellite-derived *Chla* of $0.08\pm0.02 \ \mu g \ l^{-1}$ representing oligotrophic waters (O); the 404 BOUSSOLE site in the Ligurian Sea with mean *Chla* of $0.36\pm0.37 \ \mu g \ l^{-1}$ representing mesotrophic 405 406 waters (M); and the AAOT coastal site in the northern Adriatic Sea with mean Chla of $1.74\pm1.40 \ \mu g \ l^{-1}$ 407 representing coastal waters moderately dominated by sediments (C). When considering all three water 408 types (see Fig. 1), $t_d L_w/L_T$ exhibits a large range of mean values spanning from approximately 0.07-0.14 at 490 nm, 0.06-0.22 at 555 nm and 0.01-0.12 at 670 nm. These differences are mostly due to site 409 dependent changes in L_w and L_A , both contributing to L_T (see Eq. 1). 410

411 As already stated in \$2.1, the following analysis assumes the uncertainties related to the atmospheric 412 correction process do not affect the determination of satellite-derived L_w because of the use of the same 413 atmospheric models and algorithms for SVC and for atmospheric correction.



Figure 1. Spectral values of $t_d L_w/L_T$ for oligotrophic (O), mesotrophic (M) and coastal (C) waters. Mean values and standard deviations σ (indicated by the vertical error bars), result from the analysis of 814, 1487 and 1045 SeaWiFS data extractions, respectively. The center-wavelengths between spectra have been shifted by ±2 nm to increase readability.

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Fig. 2 summarizes results from the application of Eq. 3 using identical spectrally independent relative 420 421 uncertainties for in situ L_w (i.e., 5%). The derived values of $u(L_T)/L_T$ exhibit a significant spectral 422 dependence and, as expected, are smaller when $t_{d}L_w/L_T$ is smaller (i.e., in correspondence with the lower values of L_w). Specifically, the lowest $u(L_T)/L_T$ are observed for mesotrophic waters with values 423 424 included in the range of approximately 0.2-0.5% in the blue-green spectral regions, and dropping below 425 0.1% at 670 nm. The values observed for the oligotrophic waters are higher in the blue spectral region 426 with values approaching 0.7%. In agreement with the higher values of L_w , $u(L_T)/L_T$ computed for the 427 coastal waters reach 1.1% at 555 nm and 0.6% at 670 nm. It is mentioned that differences in the observation conditions at the various sites or in the spectral values of $u(L_w)/L_w$, may lead to $u(L_T)/L_T$ 428 429 different from those presented in Fig. 2. Additionally, the relative combined uncertainty value of $L_{\rm T}$ 430 determined from a number N of in situ L_w data obtained with equivalent observation conditions would

- 431 decrease with respect to the value of $u(L_T)/L_T$ from an individual L_w due to the statistical averaging of the
- 432 random component of uncertainties.



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Figure 2. Relative uncertainties $u(L_T)/L_T$ determined assuming a spectrally independent 5% uncertainty value for L_w with the mean values of $t_d \cdot L_w/L_T$ given in Fig. 1 for different water types: oligotrophic (O), mesotrophic (M) and coastal (C). The vertical bars refer to values determined with $t_d \cdot L_w/L_T \pm \sigma$.

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Figure 3. Relative uncertainties $u(L_w)/L_w$ determined assuming a spectrally independent 0.3% uncertainty value for L_T and the mean values of $t_d \cdot L_w/L_T$ given in Fig. 1 for different water types: oligotrophic (O), mesotrophic (M) and coastal (C). The vertical bars refer to values determined with $t_d \cdot L_w/L_T \pm \sigma$.

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444 Rearranging Eq. 3, relative uncertainties in satellite-derived L_{w} can be investigated as a function of 445 $u(L_{\rm T})/L_{\rm T}$. By assigning a spectrally independent value of 0.3% to $u(L_{\rm T})/L_{\rm T}$ (i.e., a value that occurs often 446 for $|\Delta g|$ in Table 3), results displayed in Fig. 3 indicate that the 5% uncertainty requirement in satellite-447 derived L_w generally cannot be met in the red for oligotrophic and mesotrophic waters, and is 448 challenging in the blue mostly at 412 nm for mesotrophic and coastal waters. Because of this, the 0.3% 449 value assigned to $u(L_T)/L_T$, could be considered a rough upper threshold for the uncertainties of g-factors 450 allowing to meet the 5% science requirement for $u(L_w)/L_w$ in the blue-green spectral regions. The same $u(L_w)/L_w$ values displayed in Fig. 3 also indicate that the application to different missions of g-factors 451 452 determined with independent in situ data sources and exhibiting typical differences of 0.3% in the blue-453 green spectral regions with respect to the values obtained with an identical in situ data source, may introduce mission dependent biases of several percent in multi-mission CDRs. These biases would 454 455 hinder stability requirements in satellite-derived products even when applying the same atmospheric 456 correction code to the processing of data from different missions. This result is confirmed by practical assessments presented in Werdell et al (2007) showing that for deep waters $\Delta g \sim 0.3\%$ may lead to biases 457 of 4% in $L_{\rm w}$ at 555 nm. 458

In addition, the spectral differences affecting the values of Δg from the same data source or across data sources (see Table 3), may lead to significant spectral inconsistencies in CDRs. These inconsistencies (i.e., substantial inter-band spectral changes of Δg) would affect the capability of meeting the 1% interband uncertainty for L_w included in some mission objectives and likely the 3% stability requirement for an ECV like *Chla* (WMO 2011), which is commonly derived from spectral ratios of L_w . A statistical index that can be of interest to discuss stability requirements for the construction of CDRs from different satellite missions, is the relative standard error of the mean (*RSEM*) of *g*-factors *g* determined from

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$$RSEM = (\sigma_a/g)/\sqrt{N_v}$$
(5)

468 with σ_g standard deviation of *g* assumed invariant with time for each considered data source, and N_y 469 the scaled number of match-ups per decade (i.e., $N_y=10\cdot N/Y$ where *N* is the number of actual matchups 470 and *Y* the number of measurement years).

The scaling of the number of matchups over a decade, that forces the assumption of continuous availability of measurements for each *in situ* data source during the considered period, is only applied to facilitate the comparability of *RSEM* values for data which were available for a limited number of years at the time of this study. Nevertheless, continuous operation and delivery of measurements are required for any *in situ* SVC data source contributing to the creation of CDRs,

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Figure 4. Plot of the standard percent error of the mean (*RSEM*) for the SeaWiFS *g*-factor given in
Table 2 and additionally for MERIS *g*-factors determined with BOUSSOLE data (i.e., BOUSSOLE-M).

481 In view of supporting such a discussion on stability requirements through actual numbers, Fig. 4
482 displays *RSEM* values computed using the data in Table 2.

The notably low values of RSEM determined with the MOBY and MOBY-MS data suggest high 483 484 measurement precision likely explained by very stable measurement conditions, systematic calibration 485 and characterization of field radiometers, robust quality assessment of field measurements and quality 486 control of data products. The higher *RSEM* values resulting from the other data sources are likely 487 explained by a number of factors including (but not restricted to): *i*. measurement conditions perturbed 488 by time-dependent changes in the marine and atmospheric optical properties or observation geometry; *ii*. 489 instability of the *in situ* measurement system when challenged by environmental perturbations during 490 deployments (e.g., bio-fouling) or by variable performance of radiometer systems operated during 491 successive deployments, or even by different measurement methods when considering a combined data 492 set; *iii* or a relatively small of number of matchups N_y per decade.

The large *RSEM* values determined for BOUSSOLE, which refer to field radiometric measurements performed during the early deployment phase of the buoy system, are due to large σ_g and a relatively small number of matchups. Successive improvements in quality assurance and control of the radiometric measurements have led to a great reduction of σ_g . This is shown by the BOUSSOLE-M *RSEM* values also displayed in Fig. 4, and computed applying recent σ_g of *g*-factors determined for the Medium Resolution Imaging Spectrometer (MERIS). These updated values of σ_g , which refer to a 7-year measurement period, vary between 0.006 and 0.012 with *N* ranging from 15 to 42.

Overall, the previous findings suggest that any element affecting reproducibility of measurements and observation conditions with time, and thus challenging the precision of *in situ* reference measurements, should be minimized to lessen perturbations affecting the random component of uncertainties for *g*factors and thus the stability requirement for CDRs resulting from the combination of multi-mission satellite-derived data. In addition, frequent swaps of radiometer systems exhibiting similar measurement uncertainties should be considered an important best practice. In fact, the measurement uncertainties would average over the number of deployments occurring during each satellite mission. This is expected to increase the probability of achieving equivalent precision for g-factors applicable to the processing of satellite data from independent missions.

To conclude, the 0.5% stability requirement over a decade (WMO 2011) entails maximum 509 510 uncertainties of approximately 0.05, 0.025 and 0.005% in g-factors, assuming generic values of 0.10, 0.05 and 0.01 for the term $t_d L_w / L_T$. These uncertainties are comparable to the RSEM values determined 511 512 for MOBY in the blue-green spectral regions over a period of approximately 10 years, while they are significantly lower than those determined from the other in situ data sources (see Fig. 4). This result 513 514 further indicates: *i*. the need for long-term highly consistent *in situ* data applicable to SVC in view of minimizing any appreciable perturbation that may affect the determination of g-factors over time for 515 516 different or successive satellite missions; and *ii*. caution in using data from sole or multiple sources, 517 which may refer to measurement conditions difficult to reproduce for different missions.

518 Additionally, the application of mission-independent atmospheric models and algorithms for the 519 atmospheric correction process is critical.

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5. Summary and Recommendations

523 SVC does not literally lead to the absolute radiometric calibration of the satellite sensor. Rather, 524 assuming equivalent observation conditions characterizing both SVC and atmospheric correction 525 processes, SVC forces the determination of satellite-derived $L_{\rm w}$ with an uncertainty comparable to that of the *in situ* reference L_w applied for the indirect calibration process. This is achieved through vicarious 526 527 adjustment gain-factors (i.e., g-factors), which are applied to the top of atmosphere radiances $L_{\rm T}$ after full instrument calibration (e.g., following pre-launch absolute calibration and characterization, and 528 529 additionally, corrections for temporal changes in radiometric sensitivity as determined through the 530 sensor-specific on-orbit calibration system).

531 The investigation presented in this work highlights that the relative uncertainty that may affect g-532 factors, to a first approximation depends on the term $t_d L_w/L_T$ and on the uncertainties affecting in situ L_w data. This finding and differences among g-factors determined for the SeaWiFS spectral bands using 533 534 various data sources, but relying on the same atmospheric models and atmospheric correction 535 algorithms, provide suggestions on the suitability of in situ L_w data sources for SVC devoted to support the construction of CDRs. Specifically, when considering the blue and green center-wavelengths 536 537 commonly applied for the determination of Chla, satellite-derived L_w resulting from the application of gfactors differing by as little as 0.3% can result in spectral biases close to 5%. These biases are several 538 539 times higher than the 0.5% target stability value per decade indicated for satellite ocean color data 540 products expected to contribute to CDRs. Thus, in view of avoiding inconsistencies in long-term data 541 records resulting from the combination of satellite products from multiple missions, a careful evaluation 542 of sites and *in situ* measurements supporting SVC is needed. In particular, the determination of g-factors 543 by combining match-ups from multiple sites, which is often a viable solution to shorten the otherwise 544 long time needed to accumulate a relatively large number of matchups satisfying early mission needs or 545 mission-specific objectives, has to be regarded as a potential source of artifacts for CDRs. In fact, even 546 assuming equivalent uncertainties for in situ data from different sources and a single atmospheric 547 correction code, differences in g-factors may result from a diverse performance of the atmospheric 548 correction process at different sites due to differences in satellite observing geometries or marine and 549 atmospheric optical properties. Further, differences in the performance of various in situ radiometer 550 systems may also affect the accuracy and precision of g-factors through those of the in situ $L_{\rm w}$ data and 551 thus also affect the stability requirements of CDRs.

In view of defining strategies for the upcoming satellite ocean color missions, the previous findings and considerations suggest that the creation of ocean color CDRs should ideally rely on: *i*. one main long-term *in situ* calibration system (site and radiometry) established and sustained with the objective to maximize accuracy and precision over time of *g*-factors and thus minimize possible biases among satellite data products from different missions; and *ii*. unique (i.e., standardized) atmospheric models and algorithms for atmospheric corrections to maximize cross-mission consistency of data products at
locations different from that supporting SVC.

559 Accounting for results from this study and any element already provided in literature, it is expected 560 that an ideal ocean color SVC site should meet the following general requirements:

Located in a region chosen to maximize the number of high-quality matchups by trading off factors
 such as best viewing geometry, sun-glint avoidance, low cloudiness, and additionally set away from
 any continental contamination and at a distance from the mainland to safely exclude any adjacency
 effect in satellite data:

2. Exhibiting known or accurately modeled optical properties coinciding with maritime atmosphere and
oligotrophic/mesotrophic waters, to represent the majority of world oceans and minimize relative
uncertainties in computed *g*-factors;

568 3. Characterized by high spatial homogeneity and small environmental variability, of both atmosphere569 and ocean, to increase precision of computed *g*-factors.

570 Any field radiometer system supporting SVC should rely on advanced in situ measurement 571 technologies, data reduction methods and quality assurance/control schemes to minimize relative 572 standard uncertainties in in situ L_w to within state-of-the-art values. In particular, uncertainty target 573 values should be 3-4% in the blue-green spectral regions and, even though not relevant for GCOS, tentatively below 5% in the red, with inter-band uncertainties lower than 1% . In particular, accounting 574 for findings from this study and from literature and without advocating the adoption of any existing SVC 575 576 radiometry system, the fulfillment of the following wide-range requirements for *in situ* radiometric 577 measurements should be considered of utmost importance:

i. Hyperspectral field data with sub-nanometer resolution to allow system vicarious calibration of any
satellite ocean color sensor regardless of its center-wavelengths and spectral responses, and thus
ensure minimization of inter-band uncertainties;

ii. State-of-the-art absolute calibration traceable to National Metrology Institutes (i.e., tentatively with
target standard calibration uncertainty lower than 2% for radiance and stability better than 0.5% per
deployment) and comprehensive characterizations of radiometers in terms of linearity, temperature
dependence, polarization sensitivity and stray light effects, in view of minimizing measurement
uncertainties and allowing for accurate determinations of uncertainty budgets;

iii. Application of quality assurance/control schemes minimizing effects of measurement perturbations
like those (when applicable) due to infrastructure shading, radiometer self-shading, wave
perturbations, bio-fouling, and additionally scheduling regular checks of *in situ* systems and frequent
swap of radiometers, as best practice to maximize long-term accuracy and precision of *in situ*reference radiometric data;

iv. Data rate ensuring generation of matchups for any satellite ocean color mission with time differences
appropriate to minimize variations in bi-directional effects due to changes in sun zenith and daily
fluctuations in the vertical distribution of phytoplankton.

In addition to requirements for establishing an ideal SVC site and generating *in situ* radiometric data with the needed accuracy and precision, the supplementary capability of continuously characterizing both the atmospheric (e.g., τ_a) and water (e.g., inherent) optical properties would provide additional important elements for the quality assurance of matchups applicable to determine *g*-factors.

It is reminded that strategies for the construction of CDRs also suggest establishing and maintaining secondary *in situ* long-term systems with performance equivalent to the main one in terms of data accuracy, precision and measurement conditions. This recommendation is enforced by the fundamental need to allow for redundancy ensuring fault-tolerance to SVC and additionally to provide optimal means for continuous verification and validation of satellite primary data products including the capability to accurately investigate systematic effects induced by different observation conditions (i.e., viewing and illumination geometry, atmosphere and water types).

It is finally mentioned that the need to standardize the atmospheric correction process for multimission data contributing to CDRs is a requirement as relevant as the availability of *in situ* data from one 607 ideal SVC site. This operational need, however, should not be seen as an impediment to further advance

atmospheric models and atmospheric correction algorithms.

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- 610

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769 Figure 1. Spectral values of  $t_d L_w/L_T$  for oligotrophic (O), mesotrophic (M) and coastal (C) waters. 770 Mean values and standard deviations  $\sigma$  (indicated by the vertical error bars), result from the analysis of 814, 1487 and 1045 SeaWiFS data extractions, respectively. The center-wavelengths between spectra 771 772 have been shifted by  $\pm 2$  nm to increase readability. 773 774 Figure 2. Relative uncertainties  $u(L_T)/L_T$  determined assuming a spectrally independent 5% uncertainty 775 value for  $L_w$  with the mean values of  $t_d L_w/L_T$  given in Fig. 1 for different water types: oligotrophic (O), 776 mesotrophic (M) and coastal (C). The vertical bars refer to values determined with  $t_d L_w/L_T \pm \sigma$ . 777 Figure 3. Relative uncertainties  $u(L_w)/L_w$  determined assuming a spectrally independent 0.3% uncertainty 778 779 value for  $L_{\rm T}$  and the mean values of  $t_{\rm d} L_{\rm w}/L_{\rm T}$  given in Fig. 1 for different water types: oligotrophic (O), mesotrophic (M) and coastal (C). The vertical bars refer to values determined with  $t_d L_w/L_T \pm \sigma$ . 780

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- Figure 4. Plot of the standard percent error of the mean (*RSEM*) for the SeaWiFS g-factor given in
- Table 2 and additionally for MERIS *g*-factors determined with BOUSSOLE data (i.e., BOUSSOLE-M).







