# **Self-shading associated with an SBA system for measurement of water-leaving radiance and its correction**

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The shading error associated with the water-leaving radiance  $(L_{w}^{V})$  measured via the Skylight **Blocked Approach (SBA, Lee** *et al.* **2013) is characterized by Monte Carlo simulations, and it is found this error is in a range of ~1-20% under most water properties and solar positions. A model for estimating this shading error is further developed, and eventually a scheme to correct this error based on the shaded measurements is proposed and evaluated. It is found that the shade-corrected value in the visible domain is within 3% of the true value, which thus indicates**  that with the SBA scheme, we can obtain not only high precision, but also high accuracy  $L_{w}^{\quad}$  in the

**field.** 

*OCIS codes: (280.0280) Remote sensing and sensors; (280.1350) Backscattering; (280.4788) Optical sensing and sensors;* 

# **1. Introduction**

Remote-sensing reflectance  $(R_{rs}$ , sr<sup>-1</sup>), defined as the ratio of water-leaving radiance  $(L_{w}$ , W m<sup>-2</sup> sr<sup>-1</sup> nm<sup>-1</sup>) to downwelling irradiance just above the surface  $(E_a$ , W m<sup>-2</sup> nm<sup>-1</sup>), is a key property in optical oceanography, from which a wide range of physical and biogeochemical properties are derived [1]. Separately, the validation of airborne or spaceborne systems for ocean color remote sensing requires accurate measurements of  $R_{rs}$  (or  $L_{w}$ ) in the field. To achieve this, several approaches have been implemented in the past decades, and the advantages and drawbacks of these are summarized in Mueller *et al.* [2] and Lee *et al.* [3]. These conventional methods generally not measure  $L_{w}$  directly, rather measure some key components and then calculate the desired  $L_{w}$ . The involved postmeasurement data processing include removal of surface reflected light, propagation of  $L_{\mu}(z)$  (vertical profile of upwelling radiance) to  $L_w$ , etc., which will bring in considerable uncertainties in the derived  $L_w$  and then  $R_{rs}$ . The Skylight-Blocked Approach (SBA, Lee *et al.* [3,4]) is a scheme to measure  $L_w$  directly, which avoids challenges in dealing with moving clouds or stratified waters and results in  $L_{w}$  with high precision. However, because the radiance sensor looks down while illumination is from above, inevitably there will be shadows from the sensor as well as from the supporting platform [5], which should be corrected for accurate  $L_{w}$  and then  $R_{rs}$ .

The shading effect (shading error) in  $L_{w}$  measurement was first discussed in Gordon and Ding [5] (represented as GD92 hereafter in short) for a sensor floating right on the surface. Based on GD92, relative shading error (*ε*) is defined as

$$
\varepsilon = \frac{L_u^{true} - L_u^{shape}}{L_u^{true}}.\tag{1}
$$

 $L_{u}^{shaded}$  and  $L_{u}^{true}$  are the upwelling radiance just beneath the surface with shading and without shading, respectively. After the transmission through the water-air interface, Eq. (1) could be rewritten as:

$$
\varepsilon = \frac{L_w^{true} - L_w^{measured}}{L_w^{true}},\tag{2}
$$

with  $L_w^{shape}$  and  $L_w^{true}$  are the water-leaving radiance with shading and without shading, respectively.

Gordon and Ding [5] pointed out that the shading error is a function of the absorption coefficient, the size of the radiometer or the housing, and solar zenith angle at subsurface. They further proposed a simple equation to relate the shading error with the above listed variables as:

$$
\varepsilon = 1 - \exp[-\frac{2aR}{\tan(\theta_w)}].
$$
\n(3)

In Eq. (3),  $\theta_{\text{w}}$  is the subsurface zenith angle of sunlight (accordingly,  $\theta_{0}$  is defined as the zenith angle of sun above surface hereafter),  $a$  (m<sup>-1</sup>) is the absorption coefficient and  $R$  (m) is the radius of the sensor or housing. In the years following the studies of Gordon and Ding (1992), shading effects for upwelling irradiance and for sensors 10's of centimeters below the surface were assessed using Monte Carlo (MC) simulations as well [5-9].

SBA is also a floating system different with that evaluated in GD92 [5] and a detailed cartoon of the sensors

and supporting apparatus is presented in Fig. 1.



Fig. 1. The SBA system (Lee et al., 2013).

It includes a buoyed platform on the center with radiance and irradiance radiometers attached on both sides. A cone (blue part in Fig. 1) is placed in the front of the radiance radiometer to block surface-reflected light.. Unlike the system addressed by GD92, however, the radiometer employed in an SBA is in air, while it is the cone used to block surface reflected light inserted  $\sim$  5 cm below the surface. The shading error associated with the SBA system was not adequately addressed, and a first-order correction following GD92 was used in Lee et al. [3]. Therefore this study aims to (1) characterize the shading error  $(\varepsilon)$  associated with the SBA system using MC simulations; and (2) develop an applicable scheme to correct the shading error associated the  $L_{\nu}$  measurement by an SBA system.

To achieve this, similarly as what was applied in some earlier studies [5,8-10], a backward MC code tracking photon events in waters was developed. Subsequently,  $L_w^{true}$  and  $L_w^{shape}$  of waters with different optical properties and sun angles were simulated, and  $\varepsilon$  was further calculated following Eq. (2). After analyzing these  $\varepsilon$  values, a revised model based on Eq. (3) is developed to express  $\varepsilon$  as a function of water properties, cone size as well as sun angle. More importantly, an effective scheme is developed to correct this shading error simply from the measured shaded  $L_n(R_n)$  spectrum while the pre-knowledge of in-situ IOPs is not required.

#### **Monte Carlo approach** 2.

The radiative transfer model involved in the MC simulation is detailed in Leathers et al. [10]. For each MC simulation of the light field, with and without the presence of the SBA system, it involves the following components and properties:

(1) Optical properties of the water: absorption and scattering coefficients of pure seawater are taken from Lee et al. [11], Pope and Fry [12], and Morel [13], respectively. For the IOPs of particles, the absorption and the scattering coefficients were simulated following the "Case-2" model embedded in Hydrolight [14-17] with the concentrations of chlorophyll [Chl, mg  $m<sup>3</sup>$ ] and suspended minerals [SPM, g  $m<sup>3</sup>$ ] as free variables. Contribution of Colored Dissolved Organic Matter (CDOM) is considered co-vary with Chl following Eq.

Simulations with a series of Chl and SPM values (as presented in Table 1) were carried out.





- (2) Scattering phase function: three scattering constituents are considered (water molecules, Chl, and SPM). The scattering phase functions of Chl and SPM based on the backscattering ratio of particles [18] were employed. The total scattering phase function is the combined scattering phase function of water molecules, Chl, and SPM.
- (3) Downwelling irradiance and light distribution: The total downwelling irradiance above surface  $(E_1(0+))$  is set as  $1.0 \text{ W nm}^{-1} \text{ m}^2$ . For indirect irradiance, it is assumed that the sky light distribution is isotropic, i.e. idealized sky condition for which the distribution of sky light is independent of the position of sun.
- (4) The ratio of indirect irradiance over total irradiance  $(r_{\rm sky})$  is specified in MC input. The  $2\pi$  space of sky is divided into 217 grids (9 zenith \* 24 azimuth + 1 polar grid). After that, radiance in each grid was determined based on  $r_{sky}$  and solar position.
- (5) Water-air interface: As discussed in Leathers et al. [8], the overall impact from wavy surface on shading error is sufficiently small, therefore a flat air-sea interface was considered in this study.

The MC code was written and compiled using Intel Fortran 95 compiler. The MC simulations applied in this effort were backward simulation with its higher efficiency than forward MC simulations. The reliability and confidence of this MC code were evaluated by comparing its radiance outputs under different viewing angle with those from Hydrolight for identical setups when there are no measurement systems, with results showing in Fig. 2.



Fig. 2. Comparison between MC results and Hydrolight.

It is found that the difference between radiances from this MC code and that from Hydrolight is generally  $\lt$  1%, which suggests that the radiance field is reliable from this MC simulation system.

In MC simulations, a shading event is considered when photons hit any part of the sensor or the platform, and the contribution of these photons will be set to 0 when calculating  $L_{\omega}^{shade}$ .

A further evaluation was carried out to estimate  $\varepsilon$  of a cylindrical housing sensor on the surface as described in GD92. It is found that the  $\varepsilon$  values between these two MC simulations for the same  $\omega$ <sub>0</sub> (single scattering albedo) agree with each other very well (less than 5% relative difference) for  $aR < 0.1$  ( $R = 0.05$  m) (see Fig. 3).



Fig. 3. Comparison of  $\varepsilon$  between MC results in this work and that in Gordon and Ding when  $\theta_0 = 30^\circ$ ,  $R = 0.05$  m. The dash line represent the 10% bias of result from MC in simulation in Gordon and Ding

For  $aR > 0.1$ , the impact of (back)scattering effect on shading error becomes more significant and higher difference is observed. But the relative difference is still within 10% ( $aR > 0.1$ ), which provides a validation of the MC code developed here. Note that due to different computer architecture and input setups (i.e., digitization of scattering phase function), there are inherently a few percent of differences in numerically simulated radiance [19], which could explain the higher difference for  $aR > 0.1$ .

After the MC code was well validated, it was then applied to the present SBA system. To easily represent the structure of the SBA system in our MC code, this complicated structure is decomposed and replaced with a combination of cylinders (the cone and two radiometers are redrew as a group of cylinders) and cuboids (i.e., supporting platform and rest of the structure) with their measures presented in Fig. 1.

#### **Results** 3.

## 3.1 Shading effect of the SBA system

Values of  $L_{\mu}$  (nadir view) with and without shading of the SBA system were simulated with the MC code for various sun angles and water properties. The shading error due to this SBA system, which is calculated following Eq.  $(2)$ , can be summarized as follow.

 $(1)$ As demonstrated in GD92, there is a strong dependence of  $\varepsilon$  on sun zenith angle  $(\theta_0)$  (see Fig. 4).  $\varepsilon$ increases rapidly from ~10% to more than 40% for  $\theta$ <sub>0</sub> varied from 60° to 0° (for  $a = 0.5$  m<sup>-1</sup>,  $R =$  $0.05$  m).



Fig. 4. Shading error of SBA for different solar position.

- $(2)$ It appears that the azimuth effect is negligible as long as the radiance sensor of the SBA system is kept with an azimuth angle  $\leq 120^\circ$  (i.e., keep the radiance sensor in the sun side) (see Fig. 4).
- $(3)$ The simulation study of GD92 focused on cases dominated by absorption coefficient, where the effect of (back)scattering on this shading effect was omitted. It is found that (back)scattering does have considerable impact on this shading, although it is likely in general secondary. For example, for  $a = 0.5$  m<sup>-1</sup>,  $\varepsilon$  changes from 15.7% with  $b_n/a = 0.15$  to 19.5% for  $b_n/a = 0.30$  (see Fig. 5), which is  $\sim$ 20% increase in shading error.



Fig. 5. Variation of shading error with a and  $b_{\mu}$ ; here  $b = 3 \times a$ .

 $(4)$ 

Besides, as presented in Fig. 6, for different  $\theta_{\rho}$ ,  $\varepsilon$  does not monotonously response to an increase of  $b_{\mu}$ . Shading error decreases with the increase  $b_{\mu}$  for low  $b_{\mu}$  values, then increases with an increase  $b<sub>k</sub>$ . This appears to be a result of the double effects of (back)scattering: a). Water in daytime does not appear as black is simply due to scattering. Therefore, for water with a given absorption coefficient, adding scattering will basically light up the water environment, thus increase the possibility of scattered photons to light up the shaded area by the cone, thus effectively reduce (although might be just slightly) the shading error. b). Backscattering at the same time functions similarly as absorption and effectively reduces the efficiency of photons going forward. Therefore, when  $b<sub>h</sub>$  increases further, this attenuation effect of light exceeds the illumination gains from backscattered photons and the shading error is increased.



Fig. 6. impact of bb on shading error ( $\varepsilon$ ); a =1 (m-1), b = 8 x a (m-1)

# 3.2 Modeling the shading error of the SBA system

As a rule of thumb Gordon and Ding [5] described that the shading error is generally a function of waters' absorption coefficient, where the effect of (back)scattering is omitted. This is applicable for waters or wavelengths where absorption coefficient dominates. However, for waters or wavelengths where scattering are significant due to extremely high concentration of Chl and/or SPM, such as many coastal and inland waters, it is necessary to include the effect due to scattering. We therefore developed a more general model aimed for applications for both clean oceanic water and coastal waters with strong (back) scattering effect.

Following GD92, a general formula for the shading error is described as:

$$
\varepsilon = 1 - \exp\left[-\left(K_{L_u} + K_L\right) \frac{R}{\tan(\theta_w)}\right],\tag{5}
$$

where  $K_{L_n}$  (m<sup>-1</sup>) is the attenuation coefficient of upwelling radiance (nadir view or zenith going) under no shade and  $K$ ,  $(m^{-1})$  is the attenuation coefficient for upwelling radiance in the shade. In this study, the radius of the cone (as presented in Fig. 1, about  $0.05$  m) is taken to be the value of R.

Following descriptions of the diffuse attenuation coefficient of downwelling irradiance [20], we may consider both  $K_{L_n}$  and  $K_L$  as functions of a and  $b_n$  as well as solar zenith angle at subsurface, and represent the sum of  $K_{L}$  and  $K_{L}$  as K. K can then in general be described as:

$$
K = f_1(a, b_b, \theta_w). \tag{6}
$$

Further, by least-square fitting of the  $\varepsilon$  data obtained from MC simulations for the wide range of IOPs and three sun angles, we obtained an empirical formula for  $K$  as:

$$
K(\lambda) = [3.15 \sin(\theta_w) + 1.15]e^{-1.57b_b(\lambda)} a(\lambda)
$$
  
+ [5.62 sin(\theta\_w) - 0.23]e^{-0.5a(\lambda)}b\_b(\lambda). (7)

The estimated  $\varepsilon$  based on the above K model for  $\theta_0 = 10^{\circ}$ , 30° and 60° is compared with that simulated by MC (see Fig. 7), where an overall average error of 5.8% is obtained. Also shown is the estimation of  $\varepsilon$  based on the formula of Gordon and Ding [5], where an overall average error of 38.1% is found for this dataset, likely a result of omitting the effect of (back)scattering in the model of GD92.



**Fig. 7.** Estimated ε through K (Eqs. (5) and (7)) and Gordon and Ding (y-axis) .VS. calculated ε from MC (x-axis).

Note that in the above evaluations and later on shading error corrections, the azimuth angle of the radiance sensor  $(\phi)$  is set at 0° (as the coordinate system present in Fig. 4), since the impact of azimuth angle on the shading effect (as long as  $\phi$  is kept < 120° from the sun plane) is negligible.

# **3.3 Correction of shading error**

# **3.3.1 Overall scheme of shading correction**

An imperative step in the  $L_{w}$  measurement by SBA (and other in-water measurements) is to correct the shading error due to the system. As shown in Gordon and Ding's study [5] and the above, this error depends on IOPs, sun angle as well as the size of the radiometer. Although the latter two can be known for any given measurement system and time and location, the IOPs are not handily available at the time of measuring  $L_w$ . Here we present an effective scheme to derive IOPs from the shaded  $L_w$ , and then use this IOP products to correct the shading error.

Let's define a remote-sensing reflectance under the shading effect as

$$
R_{rs}^{\quad} \qquad = \frac{L_w^{shade}}{E_d(0+)}.\tag{8}
$$

This  $R_{rs}^{shade}$  can then be related to the desired no-shade  $R_{rs}$  ( $R_{rs}^{true}$ ) as:

$$
R_{rs}^{\text{}\} = R_{rs}^{\text{true}}(1 - \varepsilon). \tag{9}
$$

Decades of studies [21,22] have found that  $R_{rs}^{true}$  can be modeled as a function of IOPs through

$$
R_{rs}^{true}(\lambda) = \frac{0.52r_{rs}(\lambda)}{1 - 1.7r_{rs}(\lambda)},\tag{10}
$$

with  $r_{rs}$  the remote sensing reflectance right below the surface and can be expressed as that shown in Lee *et al.* [23],

$$
r_{rs}(\lambda) = g_w \frac{b_{bw}(\lambda)}{a(\lambda) + b_b(\lambda)} + g_p \frac{b_{bp}(\lambda)}{a(\lambda) + b_b(\lambda)},
$$
\n(11)

$$
g_p(\lambda) = G_0 \{ 1 - G_1 \exp \left[ -G_2 \frac{b_p(\lambda)}{a(\lambda) + b_p(\lambda)} \right] \}.
$$
 (12)

Here  $g_w$  and  $g_p$  are two model parameters for molecular scattering and particle scattering phase functions, respectively, while values of  $G_p$ ,  $G_p$ , and  $G_2$  are constants for given light geometry and particle phase function [22]. Since  $\varepsilon$  is also a function of the absorption and backscattering coefficients (Eqs. (5)-(7)), the above models indicate that  $R_{rs}^{shade}$  is simply also a function of IOPs, therefore we may derive these IOPs from  $R_{rs}^{shade}$  spectrum similarly as that to derive IOPs from no-shade  $R_{\dots}$ .

We adopted the hyperspectral optimization processing exemplar (HOPE) described in Lee *et al*. [24] for this derivation. Briefly,  $a$  and  $b<sub>b</sub>$  are modeled as:

$$
a(\lambda) = a_w(\lambda) + a_{ph}(\lambda) + a_{dg}(\lambda),
$$
\n(13)

$$
b_b(\lambda) = b_{bw}(\lambda) + b_{bp}(\lambda). \tag{14}
$$

Here  $a_{w}$  and  $b_{bw}$  are the absorption and backscattering coefficient of pure seawater with same values used for MC simulation in this study.  $a_{ph}$  and  $a_{dp}$  are the absorption coefficient of phytoplankton and detritus and CDOM, respectively;  $b_{bp}$  is the back scattering coefficient of particles.  $a_{ph}$ ,  $a_{dg}$  and  $b_{bp}$  spectra are further modeled, respectively, as [24-26]:

$$
a_{ph}(\lambda) = [a_0(\lambda) + a_1(\lambda)\ln(P)]P,
$$
\n(15)

$$
a_{dg}(\lambda) = G \exp[S_{dg}(440 - \lambda)], \qquad (16)
$$

$$
b_{b_p}(\lambda) = \left(\frac{440}{\lambda}\right)^n b_{b_p}(440),\tag{17}
$$

In which,  $a_{\rho}$  and  $a_{\rho}$  are model constants presented in Lee [27],  $S_{dg}$  is the slope of  $a_{dg}$ . *P*, *G*, *n* and  $b_{bp}$ (440), which represent  $a_{ph}$ (440),  $a_{dg}$ (440), slope of  $b_{bp}$  and particle backscattering at 440 nm, are four free variables to model  $R_{rs}^{shade}$ . To improve the performance of this scheme, an initial guess of  $\eta(\eta_{in})$  was determined as in QAA version 6 (QAA\_v6) [28] without considering shading error at both 440 nm and 555 nm:

$$
\eta_{ini} = 2.0\{1 - 1.2 \exp[-0.9 \frac{r_{rs}(440)}{r_{rs}(555)}]\}.
$$
\n(18)

$$
\tau_{rs}(\lambda) = \frac{R_{rs}^{stable}(\lambda)}{0.52 + 1.7R_{rs}^{shared}(\lambda)}.
$$
\n(19)

Where  $R$ <sub>rs</sub> shade is the in-situ  $R$ <sub>rs</sub> (or  $R$ <sub>rs</sub> shade from MC simulation). The upper and lower boundary of variation range of  $\eta$  is set as  $0.5^* \eta_{\text{ini}}$  and  $1.5^* \eta_{\text{ini}}$ , respectively.

To estimate the value of *S dg*, a referenced band at 555 nm is chosen with its relative low shading error (normally less than 10%) and strong signal for most cases. Based on Eq. (19), neglect of shading error will result less than 10% error on estimation of  $r_{rs}$ (555), which has limited impact on estimation of  $S_{dq}$  (about 5%). With this neglect of shading error at 555nm, from Eq. (19), we have value of  $r_{rs}$ (555) calculated as:

$$
r_{rs}(555) = \frac{R_{rs}^{shade}(555)}{0.52 + 1.7R_{rs}^{shade}(555)}.
$$
\n(20)

While at the same time, from Eq.  $(11)$ ,  $r_{rs}$ (555) is a function of IOPs at 555nm as well:

$$
r_{rs}(555) = g_w \frac{b_{bw}(555)}{a(555) + b_b(555)} + g_p \frac{b_{bp}(555)}{a(555) + b_b(555)},
$$
\n(21)

From Eqs. (14)-(15) and (17),  $r_{rs}$ (555) could be expressed as:

$$
r_{rs}(555) = f_2(P, G, b_{bp}(440), \eta, S_{dg}).
$$
\n(22)

Since value of *r*<sub>*rs*</sub>(555) is known and calculated from Eq. (20), Eq. (22) could be rewritten as:

$$
S_{dg} = f_3(P, G, b_{bp}(440), \eta). \tag{23}
$$

Consequently,  $R_{rs}^{shade}$  becomes a function of four free variables *P*, *G*, *n* and  $b_{bp}(440)$  which could be derived numerically through spectral optimization (Huang *et al*., Werdell *et al*.) [29,30].

The error function for the spectral optimization is defined as:

$$
Err = \frac{\{average(\lambda = 400 - 750 \, nm)[R_{rs}^{shade\_mod}(\lambda) - R_{rs}^{shade}(\lambda)]\}^{2}\}^{0.5}}{\text{average}(\lambda = 400 - 750 \, nm)[R_{rs}^{shade}(\lambda)]}
$$
\n(24)

Where  $R_{rs}^{shade\_mod}$  is the modeled shading  $R_{rs}$  estimated through Eqs. (9)-(17),  $R_{rs}^{shade}$  is the in-situ  $R_{rs}$  (or  $R_{rs}^{shade}$ get from MC simulation). Subsequently *ε* was estimated based on the derived IOPs (Eqs. (5)-(7)). Further, from Eq. (9), the shading-corrected  $R_{rs}$  ( $R_{rs}^{correct}$ ) became:

$$
R_{rs}^{correct}(\lambda) = \frac{R_{rs}^{stack}(\lambda)}{(1 - \varepsilon(\lambda))}.
$$
 (25)

#### 3.3.2 **Evaluation of the shading-correction scheme**

To characterize the difference between the derived value (i.e.,  $R_{rs}^{correct}$ , derived IOPs) and the true value (i.e., *R<sub>rs</sub>*<sup>true</sup>, input IOPs in MC simulation or IOPs in real world), *NRMSE* (normalized root-mean-square error) is employed:

$$
NRMSE = \sqrt{\frac{\sum_{i=1}^{n} (\hat{y}_i - y_i)^2}{n}} \frac{1}{\bar{y}},
$$
\n(26)

where  $\hat{y}_i$  and  $y_i$  are the predicted and the true value of a property (e.g.,  $R_y$ ), respectively, and  $\bar{y}$  is the average of the true value.

Examples of  $R_{rs}^{correct}$  spectra for different Chl and SPM values obtained from the above processes, along with its comparison with  $R_{rs}^{true}$  spectra for  $\theta_0 = 10^\circ$ , 30°, and 60°, are presented in Figs. 8 and 9. It is found in general the *NRMSE* between  $R_{rs}^{correct}$  and  $R_{rs}^{true}$  is under 2% for these low and high Chl and SPM waters, and for  $\theta_{0}$  from 10° to 60°, respectively. Larger (> 15%) difference is found for wavelengths beyond 720 nm and for  $\theta_{\rho}$  as 10° (as shown in Fig. 8), a result of  $R_{ex}$  with high absorption (> 1.2 m<sup>-1</sup>) and under intense self-shading (> 40%, 28%, 20% when  $\theta_{0} = 10^{\circ}$ , 30°, 60°, respectively). For such scenarios, the accuracy of  $R_{rs}$  is not that valuable as its application in ocean color remote sensing is limited. Separately, from Fig. 8, about 10% relative difference of *ε* is found in blue bands ( $\lambda = 400 - 440$  nm) for several cases (Chl = 5.0 mg m<sup>-3</sup>, SPM = 0.0, 1.0 and 5.0 g m<sup>-3</sup> when  $\theta_{0} = 10^{\circ}$ ). These are simply due to a mismatch between the power-law model (Eq.  $(17)$ ) used to describe the  $b_{bp}$  spectrum did not match the  $b_{bp}$  spectrum used in MC simulations. Consequently, larger errors in the retrieved IOPs were resulted, which then propagates to the estimation of *ε*. Nevertheless, even with such conditions, on average the resulted  $R_{rs}^{correct}$  ( $\lambda = 400 - 720$  nm) is still within 3% (3%, 1.5%, 1% when  $\theta_{0} = 10^{\circ}$ , 30°, 60°, respectively) of  $R_{rs}^{true}$ , which suggests a successful removal of the shading effects, at least for cases studied here.



**Fig. 8.** Comparison between shaded-corrected  $R_{rs}$  and true  $R_{rs}$  generated from MC.

Blue: true  $R_{rs}$  generated from MC; Red: shaded-corrected  $R_{rs}$ ; Black: shaded  $R_{rs}$  generated from MC



**Fig. 9.** Comparison between shaded-corrected  $R_{rs}$  and true  $R_{rs}$  generated from MC.

 $R_{rs}$ <sup>s</sup>: shaded  $R_{rs}$ ;  $R_{rs}$ <sup>e</sup>: shaded-corrected  $R_{rs}$ ;  $R_{rs}$ <sup>true</sup>: true  $R_{rs}$  get from MC simulation

# **4. Discussion**

#### **4.1 Im mpact of inacc curate IOPs on n the shading c correction**

The optimization scheme results in spectra of *a* and  $b<sub>b</sub>$  with least *Err* between  $R_{rs}^{shape}$  and  $R_{rs}^{shape\_mod}$ . However,

least Err for  $R_{\rm{R}}$  does not guarantee the best match between the real IOPs (here refer to the IOPs used in MC simulations) and the retrieved IOPs. As shown in Figs.  $10(a)$ ,  $10(b)$  and 11, inconsistent estimations of both a and  $b_{i}$  are found.



Fig. 10. Derived IOPs from New model .VS. Input IOPs in MC.

In this effort, the NRMSE of derived IOPs are on average about 6.6% for a and 10.6% for  $b<sub>b</sub>$ . As mentioned above, the errors of derived IOPs are mainly due to mismatches between the IOPs spectral models (Eqs.  $(15) - (17)$ ) and the IOPs spectra used in MC simulations. Table 2 presents the *NRMSE* of derived IOPs and corrected  $R_{rs}$  for four representative cases.



Much higher *NRMSE* is found for derived IOPs in the case Chl = 5.0 mg m<sup>-3</sup>, SPM = 1.0 g m<sup>-3</sup>. For this case, a significant mismatch in spectral shape of backscattering coefficient is observed in the green bands (490-600 nm) (Fig. 11) and result an underestimation of  $b_{b_n}$ . Consequently, to compensate this overestimation, higher absorption coefficient than true values will be derived as shown in Fig. 11. However, with this mismatch of derived IOPs, the NRMSE of  $R_{\mu}$  in this case remains at an excellent level (< 2%). This is because, with the current SBA physical size (radius of the cone is about 5 cm), for  $a(440) = 0.5$  m<sup>-1</sup> and  $b<sub>h</sub>(440) = 0.05$  m<sup>-1</sup> (values that cover > 90% of global oceans), a 10% error in a(440) and  $b<sub>k</sub>$ (440) will only result in less than 3%, 2%, 1% relative difference in  $R<sub>k</sub>$  for  $\theta<sub>0</sub>$  $= 10^{\circ}$ , 30°, and 60°, respectively. This indicates that, for a radius of the cone at 5 cm, for most water body in the world, the inaccuracy in derived IOPs will only bring limit impact on shading correction.



Fig. 11. Derived a and bb .VS. MC input a and  $b<sub>b</sub>$  when  $\theta<sub>0</sub> = 30^{\circ}$ MC: MC input; DE: Derived; Solid line: absorption; Star: backscattering

# 4.2 Sensitivity of  $\varepsilon$  to R

Based on Eq. (5),  $\varepsilon$  is a function of IOPs,  $\theta$  and R. IOPs are aquatic parameters decided by nature. Obviously (see Table 3), the smaller the  $R$ , the less the shading effect and the less error that may be brought in from shading correction. For  $\theta_{\rho}$  due to various limitation in field measurements (matching up with satellite or cruise schedule etc.), it is a parameter that cannot be pre-determined. R depends on the design of the SBA system, which is the only parameter that can be "manipulated" at the system manufacture stage. Table 3 shows the range of  $\varepsilon$  for a few value of R. For the present SBA system, for R as 5 cm, at  $\theta_0 = 30^\circ$ ,  $\varepsilon$  decreases from 11.7% to 1.2% with K decrease from 1.0 m<sup>-1</sup> to 0.1 m<sup>-1</sup>. If R could be reduced, say to 2 cm, for the same conditions,  $\varepsilon$  changes from 4.9% to 0.5% (see Table 3).

$heta_0 = 30^\circ$		$\varepsilon$ (%)	
R(m)	$K = 0.1$ (m <sup>-1</sup> )	$K = 0.5$ (m <sup>-1</sup> )	$K = 1.0$ (m <sup>-1</sup> )
0.01	0.25	1.24	2.46
0.02	0.50	2.46	4.85
0.03	0.74	3.66	7.19
0.04	0.99	4.85	9.47
0.05	1.24	6.03	11.70
0.06	1.48	7.19	13.87
0.07	1.73	8.34	15.98
0.08	1.97	9.47	18.05
0.09	2.21	10.59	20.06
0.10	2.46	11.70	22.02

Table 3. Shading error  $(\epsilon)$  under different K value and radius  $(R)$ of sensor when solar zenith  $(\theta_0) = 30^\circ$ 

In particular, if R can be kept at 3 cm or smaller, the shading error is generally less than 7.2% even for K is as high as 1.0 m<sup>-1</sup> when  $\theta_{\rho} = 10^{\circ}$  (*a* and  $b_{h}$  are about 0.5 m<sup>-1</sup> and 0.05 m<sup>-1</sup>, respectively). Also, if applying to the most turbid case (Ch1 = 5.0 mg m<sup>-3</sup>, SPM = 5.0 g m<sup>-3</sup>; corresponding  $a = 0.66$  m<sup>-1</sup> and  $b<sub>b</sub> = 0.15$  m<sup>-1</sup>) in the MC simulation as another example, for solar zenith angle varies from  $10^{\circ}$  to  $60^{\circ}$ , reducing R from 5 cm to 2 cm will cut more than half of the shading error (see Table 4). These results indicate high confidence in precisely and accurately measuring  $L_{\omega}$  in the field in the 350-720 nm range if an SBA system can be manufactured with a smallsized radiometer.



Table 4. Shading error ( $\varepsilon$ ) at 440 nm for case Chl = 5.0 (mg m<sup>-3</sup>), SPM = 5.0 (g m<sup>-3</sup>) under different radius (R) of the cone

# 5. Conclusions

In ocean color remote sensing, to achieve precise and accurate measurement of  $L_n(R_n)$  in-situ has been a long elusive goal. The above-water approach runs into the difficulty to remove surface-reflected light, while the inwater systems suffer shading errors and associated with the uncertainties of manually propagating upwelling radiance to  $L_w$ . The SBA system does measure  $L_w$  directly with high precisions, but, inevitably, there are selfshading errors associated with such measurements. Here, following the approach of Gordon and Ding [5], we characterized the shading errors for various optical properties, sun angles, and size structure using Monte Carlo simulations. It is found that depends on values of these parameters, the shading error can change from negligible to more than 20% in most cases. We further revised the model of GD92 [5] to include the effects of backscattering for the shading effect. More importantly, a practical scheme is developed to correct the shading effect simply based on the measured, shaded,  $L_w(R_{rs})$ . The resulted, shading-corrected,  $R_{rs}$  is found generally within 2% of the true (without shading)  $R_{\text{ref}}$ . These results indicate that, with the SBA concept and the correction scheme, we can now not only obtain  $R_{\mu}$  with high precision, but with high accuracy. However, as indicated in Lee et al. [3], the ultimate, by-passing sophisticated post-measurement processing, solution for high-precision and high-accuracy  $R_{\nu}$ is to manufacture an SBA system with a small-sized radiometer for radiance.

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#### **Appendix: IOPs of each components for MC simulations**

Similar as the Hydrolight "Case-2" model setting, the IOPs input for MC simulations are divided into four parts: pure water, chlorophyll, CDOM and SPM.

As mentioned previously, the IOPs of pure water are taken from Lee *et al.* [11], Pope and Fry [12], and Morel [13], respectively. For the other three constituents, the same models as those embedded in Hydrolight are applied.

The IOPs of Chl is estimated by:

$$
a_{chl}(\lambda) = a^*(\lambda)[CHL]^{0.65} [17], \qquad (27)
$$

$$
b_{chl}(\lambda) = 0.3a^*(\lambda)[CHL] \left(\frac{555}{\lambda}\right)^{0.62} [15].
$$
 (28)

Where [CHL] is the concentration of chlorophyll (mg  $m<sup>-3</sup>$ ). A backscattering ratio of 0.01 is assign to chlorophyll.

For CDOM, the absorption is calculated as presented in Eq. (4).

For SPM, the IOPs is determined by:

$$
a_{SPM}(\lambda) = a^*_{SPM}(\lambda)[SPM],\tag{29}
$$

$$
b_{SPM}(\lambda) = b^*_{SPM}(\lambda)[SPM].
$$
\n(30)

Where [SPM] is the concentration of SPM in (g m<sup>-3</sup>). The value of  $a^*_{SPM}$  and  $b^*_{SPM}$  is taken from average specific particle absorption and scattering in Hydrolight, which could be retrieved from "HE5/data/defaults/astarmin\_average.txt", "HE5/data/defaults/bstarmin\_average.txt", respectively. A backscattering ratio of 0.03 is assigned to SPM.

The volume scattering function will be constructed based on the back scattering ratio [18].

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