

Article

# Estimating underwater light regime under spatially heterogeneous sea ice in the Arctic

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**Abstract:** The vertical diffuse attenuation coefficient for downward irradiance ( $K_d$ ) is an apparent optical property commonly used in primary production models to propagate incident solar radiation in the water column. In open water, estimating  $K_d$  is relatively straightforward when a vertical profile of measurements of downward irradiance,  $E_d$ , is available. In the Arctic, the ice pack is characterized by a complex mosaic composed of sea ice with snow, ridges, melt ponds and leads. Because of the resulting spatially heterogeneous light field in the first metres of the water column, it is difficult to measure at single-point locations meaningful  $K_d$  values that allow predicting average irradiance at any depth. The main objective of this work is to propose a new method to estimate average irradiance over large spatially heterogeneous area as it would be seen by drifting phytoplankton. Using both in situ data and 3D Monte Carlo numerical simulations, we show that (1) the large-area average downward irradiance profile,  $\overline{E_d}(z)$ , under heterogeneous sea ice cover can be represented by a single-term exponential function and (2) the vertical attenuation coefficient for upward radiance ( $K_{Lu}$ ), which is up to two times less influenced by a heterogeneous incident light field than  $K_d$  in the vicinity of a melt pond, can be used as a proxy to estimate  $\overline{E_d}(z)$  in the water column.

**Keywords:** apparent optical properties, 3D Monte Carlo numerical simulations, downward irradiance, upward radiance, sea ice heterogeneity, vertical attenuation coefficient, melt ponds

## 1. Introduction

The vertical distribution of underwater light is an important driver of many aquatic processes, such as primary production by phytoplankton, and photochemical reactions, such as the photodegradation of organic matter. Hence, an adequate description of the underwater light regime is mandatory to understand energy fluxes in aquatic ecosystems. In open water, when assuming an optically homogeneous water column, downward irradiance at any given wavelength follows, as a first approximation, quite well a monotonically exponential decrease with depth, which can be modelled as follows [1] (Equation 1):

$$E_d(z) = E_d(0^-) e^{-K_d(z) z} \quad (1)$$

where  $E_d(z)$  is the downward irradiance ( $\text{W m}^{-2}$ ) at depth  $z$  (m),  $E_d(0^-)$  is the downward irradiance ( $\text{W m}^{-2}$ ) just below the surface and  $K_d(z)$  is the diffuse vertical attenuation coefficient ( $\text{m}^{-1}$ ) describing the rate at which light decreases with increasing depth.  $K_d$  is one of the most commonly used apparent optical properties (AOP) of seawater, and a good estimation of this parameter is important

for measuring or modelling primary production.  $K_d$  may vary with depth because changes in inherent optical properties and/or in the structure of the light field. But as Kirk [1] pointed out, for practical considerations in oceanography and limnology, the  $K_d$  value, even averaged in the euphotic zone, is a useful and valuable way to represent the downward irradiance attenuation in that upper layer. For example, to determine primary production based on simulated on-deck incubations or photosynthetic parameters derived from photosynthesis vs. irradiance curves (P vs. E curves) requires measured or estimated values of  $K_d$  (e.g. Morel [2]). Nowadays,  $K_d$  is relatively easy to estimate using commercially available radiometers.

The ice-infested regions of the Arctic ocean are characterized by a complex mosaic made of sea ice with snow, melt ponds, ridges and leads [3–5]. Phytoplankton is exposed to a highly variable light regime while drifting under these features (e.g. Lange *et al.* [6]). Estimating primary production of phytoplankton under sea ice requires an approach that is adequate to capture this large-area variability in the light field. In situ incubations at single locations of seawater samples inoculated with  $^{14}\text{C}$  or  $^{13}\text{C}$  are not appropriate because they reflect primary production under local light conditions, which is not representative of the range of irradiance experienced by drifting phytoplankton over a large area. One classical approach that is more adequate consists in conducting on-deck simulated 24-hours incubations of seawater samples inoculated with  $^{14}\text{C}$  or  $^{13}\text{C}$  and applying the light attenuation at the depths of sample collections, using natural illumination and neutral filters. An alternative approach consists in calculating primary production using modelled or measured daily time series of incident irradiance, sea ice transmittance and in-water vertical attenuation coefficients, combined with photosynthetic parameters determined on P vs. E curves measured with short (under two hours) incubations of seawater samples inoculated with  $^{14}\text{C}$ . The latter two methods require that the vertical profile of the irradiance experienced by drifting phytoplankton be appropriately determined, which is challenging due to surface heterogeneity. Traditionally, one or very few  $E_d(z)$  profiles are measured at discrete locations under sea ice (e.g. Mundy *et al.* [7]). Such measurements, however, do not capture the variability induced by sea ice features. In recent studies, to better document the spatial variability of  $E_d(z)$ , radiometers were attached to either remotely operated vehicles (ROV) [4] or a surface and under-ice trawl (SUIT), a net developed for deployment in ice-covered waters, typically behind an icebreaker [6]. Both an ROV and a SUIT allow a better description of the light field right under sea ice, which is more appropriate for determining average irradiance experienced by drifting phytoplankton. Such under-ice measurements can then be combined with averaged  $K_d$  values to propagate light at depth.

Estimating irradiance at depth for primary production measurement or calculation using  $K_d$  values derived from only a few discrete vertical profiles of  $E_d(z)$  under heterogeneous sea ice is problematic whatever the platform for radiometer deployment. Let us consider that phytoplankton, by continuously drifting horizontally relative to sea ice, is exposed to fluctuations in irradiance due to surface heterogeneity, and that the relevant light metrics for primary production in such conditions is irradiance at any depth averaged over some horizontal area. When measuring an irradiance profile at one given location under sea ice, as the depth of the upward-looking detector increases, light from a larger area on the underside of the ice enters the detector field of view. In other words, the detector "sees" different things at different depths. One consequence is that  $E_d(z)$  measured that way may not follow the usual monotonically exponential decrease with increasing depth (Equation 1). For example, irradiance profiles measured beneath low-transmission sea ice (e.g. white ice) relative to surrounding areas showing melt ponds, show subsurface light maxima. The literature reports subsurface maxima varying between five and 15 m in depth [5,8,9]. Conversely, it is also important to note that  $K_d$  estimations are biased when profiles are measured beneath an area of high transmission (e.g. a melt pond) relative to surrounding areas [5]. Indeed, with depth, light decreases more quickly than what would be expected from the inherent optical properties (IOPs) of the water column. In the field, this situation is more difficult to identify compared to profiles showing subsurface maxima because the former measurements may appear to follow a single exponential decrease but would not

produce a diffuse attenuation coefficient that adequately describes the water mass. So, two vertical light profiles measured a few metres apart under sea ice are often very different. More importantly, local measurements of light under heterogeneous sea ice do not provide an adequate description of the average light field as it would be seen by drifting phytoplankton cells at different depths. This makes estimations of primary production and the interpretation of biogeochemical data challenging in the presence of sea ice.

To fit vertical profiles of  $E_d(z)$  under bare ice that do not follow an exponential decay under sea ice covered with melt ponds, Frey *et al.* [8] proposes a simple geometric model (Equation 2).

$$E_d(z) = \pi E_d(0^-)(1 + P(N - 1) \cos^2 \phi) e^{-K_d(z) z} \quad (2)$$

where  $E_d(0^-)$  is the irradiance directly below the ice/snow,  $P$  the areal fraction of the ice cover,  $N$  the ratio between ice and melt ponds transmittance and  $\phi$  a fitting parameter defined as  $\arctan(R/z)$  with  $R$  the radius of the ice patch and  $z$  the depth. A major drawback of this method is that additional field observations of  $N$  and  $P$  are required to adequately parameterize the model, which makes its use more difficult. To address this concern (among others), Laney *et al.* [9] proposed a semi-empirical parameterization that includes a second exponential coefficient in Equation 1 to model light decrease at the interface between the ice and ocean water at the bottom of the ice layer (Equation 3):

$$E_d(z) = E_d(0^-) e^{-K_d(z) z} - (E_d(0^-) - E_d(NS)) e^{-K_{NS}(z) z} \quad (3)$$

where  $E_d(0^-)$  is the irradiance that would be observed under homogeneous snow or ice cover,  $E_d(NS)$  is the irradiance under ice, and  $K_{NS}(z)$  describes the decrease of  $E_d(0^-)$  just under the ice layer. Both the methods by Frey *et al.* [8] and Laney *et al.* [9] make it possible to propagate local  $E_d(z)$  vertically under low transmission ice. However, these methods cannot identify and correct for inflated  $K_d$  when profiles are measured beneath an area of high transmission relative to surrounding areas. Additionally, when trying to determine primary production by phytoplankton that drift under sea ice and therefore are not static under sea ice features, what matters is the average shape of the vertical  $E_d(z)$  profile, which may possibly be predictable using a large-area  $\overline{K_d}$  as under a wavy open ocean surface [10].

In this study, using both in situ data and 3D Monte Carlo numerical simulations of radiative transfer, we show that the vertical propagation of average  $E_d(z)$ ,  $\overline{E_d}(z)$ , is reasonably well approximated by a single exponential decay with a so-called large area  $K_d$ ,  $\overline{K_d}$ , under sea ice covered in melt ponds. We further demonstrate that  $\overline{K_d}$  can be estimated from the vertical attenuation coefficient for upward radiance ( $K_{Lu}$ ) because the latter is believably less affected by local surface features of the ice cover. We implicitly assume that primary production can be adequately modeled using  $\overline{E_d}(z)$ , and we conclude that  $K_{Lu}$  is an appropriate AOP for predicting the vertical variations in  $\overline{E_d}(z)$  under sea ice.

## 2. Material and methods

### 2.1. Study site and field campaign

The field campaign was part of the GreenEdge project ([www.greenedgeproject.info](http://www.greenedgeproject.info)) which was conducted on landfast ice southeast of the Qikiqtarjuaq Island in the Baffin Bay (67.4797N, 63.7895W). The field operations took place at an ice camp where the water depth was 360 m, from April 20 to July 27, 2016 (Supplementary Fig. 1). During the sampling period, the study site experienced changes in the snow cover and landfast ice thickness of between 0-49 and 106-149 cm, respectively.

### 2.2. In situ underwater light measurements

During the campaign, a total of 83 vertical light profiles were acquired using a factory-calibrated ICE-Pro (an ice floe version of the C-OPS, or Compact-Optical Profiling System, from Biospherical Instruments Inc.) equipped with both downward plane irradiance  $E_d(z)$  ( $\text{W m}^{-2}$ ) and upward radiance

$L_u(z)$  ( $\text{W m}^{-2} \text{sr}^{-1}$ ) radiometers. The ICE-Pro system is a negatively buoyant instrument with a cylindrical shape 10 inches in diameter and is not designed for free-fall casts (as opposed to its open-water version). To perform the profiles, the frame was manually lowered into an auger hole that had been cleaned of ice chunks. Once it was underneath the ice layer, fresh clean snow was shovelled back in the hole to prevent the creation of a bright spot right on top of the sensors. Great care was taken not to pollute the hole surroundings (footsteps, water and slush spillage from the auger drilling, etc.). The operator then stepped back 50 m, while keeping the sensors right under the ice, to avoid any human shadow on top of the profile. The frame was then lowered manually at a constant descent rate of approximately  $0.3 \text{ m s}^{-1}$ . The above-surface atmospheric reference sensor was fixed on a steady tripod standing on the floe approximately 2 m above the surface and above all neighbouring ice camp features. Data processing and validation were performed using a protocol inspired by the one proposed by Smith and Baker [11] which is now used by the main space agencies. Measurements were made at 19 wavelengths: 380, 395, 412, 443, 465, 490, 510, 532, 555, 560, 589, 625, 665, 683, 694, 710, 765, 780 and 875 nm. For this study,  $E_d$  and  $L_u$  spectra were interpolated linearly between 400 and 700 nm every 10 nm. In situ diffuse attenuation coefficients ( $K$ ) for both  $E_d$  ( $K_d$ ) and  $L_u$  ( $K_{Lu}$ ) were calculated on a 5 m sliding window (10–15 m, 15–20 m, ..., 70–75 m, 75–80 m) starting at 10 m to reduce the effects of surface heterogeneity. A total of 72 044 non-linear models were calculated to estimate both  $K$  coefficients from Equation 1 ( $83 \text{ profiles} \times 14 \text{ depths} \times 31 \text{ wavelengths} \times 2 \text{ radiometric quantities}$  ( $E_d$ ,  $L_u$ )). A conservative  $R^2$  of 0.99 was used essentially to filter out noisy profiles. 42 407 models were kept for subsequent analysis.

### 2.3. 3D Monte Carlo numerical simulations

#### 2.3.1. Theory and geometry

3D numerical Monte Carlo simulation is a convenient approach for modelling the light field under spatially heterogeneous sea surfaces [5,12–14]. They are simple to understand and versatile, and incident light, IOPs and geometry can be easily changed. In this study, we used SimulO, a 3D Monte Carlo software program that simulates the propagation of light in optical instruments or in ocean waters [15]. Our objective was to simulate the propagation of sunlight underneath heterogeneous ice-covered ocean waters. Simulations were performed in an idealized ocean described by a cylinder of 120 m radius and 150 m depth (Fig. 1). The water IOPs were selected to reflect pre-bloom conditions in the green/blue spectral region ( $a = b = 0.05 \text{ m}^{-1}$ ). These typical averaged values were measured during the GreenEdge 2016 campaign using an in situ spectrophotometer (ac-s from Sea-Bird Scientific) and represent the contribution of both pure water and the water constituents. The scattering phase function was described by a Fournier-Forand analytic form with a 3% backscatter fraction [16,17]. The inclusion of a 3D sea ice layer at the upper boundary of the ocean would require extensive computing power because of the high scattering properties of sea ice. Instead, sea ice was incorporated at the upper boundary of the ocean using a 2D light-emitting surface with a radius of 100 m. The angular distribution and amplitude of the light field emitted by the surface was chosen to mimic observed field data [18]. SimulO does not allow the use of arbitrary angular distribution for photon-emitting surfaces. To overcome this problem, two sources of photons were summed up in order to reproduce an observed under-ice light field (Fig. 2). The first source was a regular Lambertian emitting surface while the second was a Lambertian emitting surface but restricted to an emission within 60 degrees of the zenith angle. A 5-m radius melt pond was set up at the centre of the emitting surface (Fig. 1). The melt pond had the same emitting angular distribution as the surrounding ice. Its intensity was four times higher than the surrounding ice, which corresponds to typical conditions found in the Arctic during summer [19].

Given our interest in surface light profiles, 2D horizontal software detectors were placed vertically every 0.5 m, from 0.5 m up to a depth of 25 m. Detectors include  $1 \text{ m}^2$  pixels measuring downward irradiance and upward radiance (5-degree half angle). In order to avoid the effect of the boundary (i.e.,

absorption by the side of the cylinder used to simulate the water column), data outside a radius of 50 m were not used (see the green box in Fig. 1). A total number of  $7.14 \times 10^{10}$  photons were simulated to obtain a sufficient number of upwelling photons. The simulation took approximately 6 000 hours distributed over 2 000 CPU cores. Since the geometry was symmetrical azimuthally, irradiance and radiance were averaged over the azimuth in order to raise the signal-to-noise ratio. Due to the low scattering coefficients used to reproduce in situ conditions observed during the sampling campaign, radiance profiles were noisy because a small number of upward photons could be captured. To address this issue, radiance profiles were smoothed using a Gaussian fit (Supplementary Fig. 2).

### 2.3.2. Estimation of reference and local light profiles

To explore how the melt pond influences the averaged underwater irradiance and radiance profiles (Fig. 1), data from the Monte Carlo simulation were averaged according to six different radii, corresponding to varying melt pond spatial proportions. The simulated light profiles were averaged within the following surface areas: (1) 10 m radius (25% melt pond cover), (2) 11.18 m radius (20% melt pond cover), (3) 12.91 m radius (15% melt pond cover), (4) 15.81 m radius (10% melt pond cover), (5) 22.36 m radius (5% melt pond cover) and (6) 50 m radius (1% melt pond cover). For each of these six configurations, the corresponding averaged light profile,  $\overline{E_d}(z)$ , was subsequently viewed as an adequate description of the average underwater light field. For the remainder of the text, these averaged profiles are referred to as reference light profiles. Furthermore, 50 light profiles, evenly spaced by 1 m from the melt pond centre, were extracted to mimic local measurements of light and to calculate associated diffuse attenuation coefficients.

### 2.4. Statistical analysis

All statistical analyses and graphics were carried out with R 3.5.1 [20].

## 3. Results

### 3.1. Comparing in situ downward irradiance ( $E_d$ ) and upward radiance ( $L_u$ ) measurements

An example showing in situ downward irradiance ( $E_d$ ) profiles and upward radiance ( $L_u$ ) profiles at 16 visible wavelengths measured under ice is presented in Fig. 3. For the  $E_d$  profiles, subsurface light maxima at a depth of around 10 m are clearly visible between 400 and 560 nm. These peaks are not visible in the yellow/red region (580–700 nm). For the  $L_u$  profiles, no subsurface light maxima were found at any wavelength. To have a closer look at the shape of both  $E_d$  and  $L_u$  light profiles, data below 10 m were normalized to the value at 10 m (Fig. 4). Below 10 m and between 400 and 580 nm, both  $E_d$  and  $L_u$  profiles presented the same shape (i.e., yield the same rate of extinction with increasing depth). At longer wavelengths ( $\geq 600$  nm), differences between the shapes of  $E_d$  and  $L_u$  profiles increased. Irradiance and radiance diffuse attenuation coefficients ( $K_d$  and  $K_{L_u}$ ) calculated on layers of a 5 m depth are compared in Fig. 5 for all 83 profiles. In the blue/green/yellow regions (400–580 nm), the determination coefficients between  $K_{L_u}$  and  $K_d$  varied between 0.98 at the surface (10–15 m) and 0.64 at depth (75–80 m). For most of the surface layers, regression lines lined up with the 1:1 lines. Slight deviations from the 1:1 lines started to appear after 60 m where  $K_d$  was on average higher than  $K_{L_u}$ . The relationships including orange and red wavelengths are presented in Supplementary Fig. 3. A linear regression analysis between all in situ normalized  $E_d$  and  $L_u$  profiles showed that determination coefficients ( $R^2$ ) range between 0.75 and 1 (Supplementary Fig. 4). A sharp decrease and a high variability of calculated  $R^2$  occurred beyond 575 nm. This suggests a gradual decoupling between  $E_d$  and  $L_u$  profiles at longer wavelengths, likely due to the effect of inelastic scattering (mostly, Raman).



### 3.2. 3D Monte Carlo numerical simulations

Fig. 6 shows cross-sections of the simulated downward irradiance and upward radiance. A key difference for the upcoming discussion is that the simulated upward radiance was more homogeneous compared to the simulated downward irradiance. Fig. 7 shows the reference irradiance,  $E_d(z)$ , and reference radiance,  $L_u(z)$ , profiles. The highest irradiance and radiance occurred when the melt pond occupied 25% of the sampling area, allowing for more light to propagate in the water column. None of the  $E_d(z)$  and  $L_u(z)$  reference profiles showed subsurface light maxima. Fig. 8 shows the 50 simulated local downward irradiance and upward radiance ~~light~~ profiles evenly spaced by 1 m from the melt pond ~~centre~~. Local downward irradiance profiles under the melt pond (0–5 m) showed a rapid decrease with increasing depth described by a monotonically exponential or quasi-exponential decrease. Local simulated downward irradiance profiles just outside the melt pond (5–10 m) were characterized with subsurface light maxima occurring at a depth of between approximately 5 and 10 m. Further away from the melt pond centre, downward irradiance profiles followed a monotonically exponential or quasi-exponential decrease. None of the simulated upward radiance ~~light~~ profiles presented subsurface light maxima (Fig. 8). From local simulated irradiance and radiance profiles (Fig. 8),  $K_d$  and  $K_{Lu}$  were calculated by fitting Equation 1 between 0 and 25 m. Results are presented in Fig. 9.  $K_d$  varied between 0.065 and 0.157  $\text{m}^{-1}$  and  $K_{Lu}$  between 0.079 and 0.116  $\text{m}^{-1}$ . These  $K_d$  and  $K_{Lu}$  were used to propagate light downward from surface reference values  $E_d(0^-)$ . Fig. 10 shows the profiles resulting from this operation. A greater dispersion around the reference profiles (thick black lines in Fig. 10) occurred when using  $K_d$  compared to the profiles generated with similarly derived  $K_{Lu}$  values. The relative differences between the depth-integrated values of each local profiles (coloured lines in Fig. 10) and the depth-integrated values of the reference profiles (thick black lines in Fig. 10) were used to quantify the error of using either  $K_d$  or  $K_{Lu}$  as a proxy to predict downward irradiance in the water column (Fig. 11). Below the melt pond,  $K_d$  overestimated the total downward irradiance by up to 40% for the 1% melt pond reference surface. In this region, the local  $K$  coefficients are inflated. In the transition region, between 5 and 10 m from the centre of the melt pond, where subsurface maxima are observed,  $K_d$  underestimated the downward irradiance by up to 35% for the 25% melt pond reference surface. Further away from the edge of the melt pond, the errors saturated to maximum -25%. The same behaviour is observed for  $K_{Lu}$  but with about two times less amplitude. The mean relative errors were lower by approximately a factor of two when using  $K_{Lu}$  (-7%) compared to  $K_d$  (-12%). Also, the prediction errors stabilized at a shorter distance from the centre of the melt pond when using  $K_{Lu}$  ( $\approx 10$  m) compared with using  $K_d$  ( $\approx 20$  m).

### 3.3. Inelastic scattering

Based on in situ data, our results have pointed out that  $K_{Lu}$  is not a good proxy for  $K_d$  at longer wavelengths (Supplementary Figs. 3–4) because of the effect of Raman scattering. To validate this hypothesis, we used the HydroLight (Sequoia Scientific, Inc.) radiative transfer numerical model to calculate theoretical downward irradiance and upward radiance and their associated vertical attenuation coefficients in an open water column in the presence of Raman scattering. The simulation was parameterized using IOPs measured during the field campaign (detailed information can be found in the supplementary section entitled Raman inelastic scattering). The simulation was able to reproduce the observed decoupling between  $K_d$  and  $K_{Lu}$  observed larger wavelengths  $\geq 600$  nm (Supplementary Fig. 5).

## 4. Discussion

In the Arctic, melt pond coverage, lead coverage, and ice and snow thickness can vary greatly in both time and space [21,22]. Due to this sea ice heterogeneity, local under-ice measurements of downward irradiance are sometimes characterized by subsurface light maxima (Fig. 3). To model such profiles, Laney *et al.* [9] proposed a semi-empirical parameterization using two exponential terms (see

Equation 3). Whereas their method might provide adequate estimations of instantaneous downward diffuse attenuation coefficients at specific locations, fitting a double exponential might not be ideal because data are modelled locally and do not provide an adequate description of the average light field ( $\overline{E_d(z)}$ ) as it would be seen, for example, by drifting phytoplankton cells. In such conditions, this paper argues that under-ice irradiance measurements should be analyzed in the context of ice and surface properties within a radius of several metres over the horizontal distance since local measurements cannot be used as a proxy of the average light field.

Using in situ light measurements, it was found that  $E_d$  and  $L_u$  (and therefore  $K_d$  and  $K_{Lu}$ ) were highly correlated below 10 m depth (Fig. 4, Fig. 5), even when subsurface light maxima were present (Fig. 3). Furthermore, no subsurface light maxima were observed in the in situ upward radiance profiles. The reason is that a  $L_u$  radiometer measures upwelling photons coming from deeper depth that have undergone more scattering. These photons thus originate from a larger surface area. This reinforces the idea that  $L_u$  is less influenced by sea ice surface heterogeneity.

Based on Monte Carlo simulations, our results showed that the average downward irradiance profile,  $\overline{E_d(z)}$ , under heterogeneous sea ice cover follows a single-term exponential function, even when melt ponds occupy a large fraction of the study area (Fig. 7). This is similar to what is observed under a wavy ice-free surface [10]. However, estimating  $\overline{E_d(z)}$  for a given area is not straightforward, as it requires a large number of local profiles under the sea ice. An intuitive alternative to deriving the attenuation coefficient is to use upward radiance, which is less influenced by sea surface heterogeneity compared to downward irradiance (Fig. 3, Fig. 4, Fig. 5). Monte Carlo simulations showed that a local estimation of  $K_{Lu}$  was a good proxy for  $\overline{K_d}$  and that using  $K_{Lu}$  rather than  $K_d$  provided better estimations of the average downward profile by reducing the average error by approximately a factor of two (Fig. 11).

There are at least two main factors influencing the quality of in situ downward irradiance measurements under heterogeneous sea ice. The first factor is the horizontal distance from the centre of the melt pond. Although the relative error of propagating  $E_d(0^-)$  using both  $K_d$  and  $K_{Lu}$  showed the same pattern, the largest error occurred when using local estimations of  $K_d$  directly below the melt pond and up to 10 m from the melt pond edge (Fig. 11). In contrast, the relative error associated with the use of  $K_{Lu}$  was much lower and stabilized just after approximately 10 m from the centre of the melt pond. The second factor driving the relative error of local measurements is the proportion occupied by melt ponds over the area of interest (Fig. 11). Indeed, higher proportions of melt pond allow for more light to penetrate in the water column. Hence, local measurements made under surrounding ice are more likely to show subsurface light maxima (see Frey *et al.* [8]). Accordingly, when melt ponds accounted for 1% of the total area, averaged error in  $E_d(z)$  using  $K_{Lu}$  was 1.33% but increased to 18% when the melt pond occupied 25% of the total area (Fig. 11).

## 5. Conclusions

Our results show that under spatially heterogeneous sea ice at the surface (and for a homogeneous water column), the average irradiance profile,  $\overline{E_d(z)}$ , is well reproduced by a single exponential function. We also showed that propagating  $E_d(0^-)$  using  $K_{Lu}$  is a better choice compared to  $K_d$  under heterogeneous sea ice. Nowadays, radiance measurements are becoming more routinely performed during field campaigns, so we argue that one should use  $K_{Lu}$  when available to propagate  $E_d(0^-)$  through the water column under sea ice. The main difficulty remains in finding good estimates of averaged  $E_d(0^-)$ . In recent years, this has become easier with the development of remotely operated vehicles [3,4,23], remote sensing techniques and drone imagery. In this study, we used a Monte Carlo approach to model an idealized surface with a single melt pond (Fig. 1, Fig. 6). Fig. 11 shows that the effect of a 5 m melt pond is minimized after approximately 20 m. Therefore, when many melt ponds are characterizing an area, if one has to perform a single profile, measuring an upward radiance profile under bare ice as far away as possible from any melt pond would minimize the error in estimating the area-averaged downward irradiance profile using  $K_{Lu}$ . Although not representative of a complex

308 Arctic sea ice surface, our simple surface geometry allowed to study the transition from a high to a  
309 low transmission sea ice. Further 3D Monte Carlo work could include a more complex geometry of  
310 heterogeneous surfaces.



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