

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 plane 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 top meters 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 of radiative transfer, we show that (1) the large-area average vertical profile of downward irradiance, $\overline{E}_d(z)$, 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 plane irradiance (W m^{-2}) at depth z (m), $E_d(0^-)$ is the downward plane irradiance (W m^{-2}) just below the surface and $K_d(z)$ is the diffuse vertical attenuation coefficient (m^{-1}) describing the rate at which downward irradiance decreases with increasing depth. K_d is one of the

29 most commonly used apparent optical properties (AOP) of seawater, and a good estimation of this
30 parameter is important for measuring or modelling primary production. K_d may vary with depth
31 because of changes in seawater inherent optical properties, the angular structure of the light field,
32 and the effects of inelastic radiative processes such as Raman scattering by water molecules and
33 fluorescence by phytoplankton pigments or dissolved organic matter. As Kirk [1] pointed out, for
34 practical considerations in oceanography and limnology, the K_d value, even when averaged within the
35 euphotic zone, provides a useful proxy to represent the downward irradiance attenuation in the upper
36 water column. For example, to determine primary production based on simulated on-deck incubations
37 or photosynthetic parameters derived from photosynthesis vs. irradiance curves (P vs. E curves)
38 requires measured or estimated values of K_d (e.g., Morel [2]). Nowadays, K_d is relatively easy to
39 estimate using commercially available radiometers.

40 The ice-infested regions of the Arctic ocean are characterized by a complex mosaic made of sea
41 ice with snow, melt ponds, ridges, and leads [3–5]. Phytoplankton are exposed to a highly variable
42 light regime while drifting under these heterogeneous features (e.g., Lange *et al.* [6]). Estimating
43 primary production of phytoplankton under sea ice requires an approach that is adequate to capture
44 this large-area variability in the light field. In situ incubations at single locations of seawater samples
45 inoculated with ^{14}C or ^{13}C are not appropriate because they reflect primary production under local light
46 conditions, which is not representative of the range of irradiance experienced by drifting phytoplankton
47 over a large area. One classical approach that is more adequate consists in conducting on-deck
48 simulated 24-hours incubations of seawater samples inoculated with ^{14}C or ^{13}C and applying the light
49 attenuation at the depths of sample collections, using natural illumination and neutral filters. An
50 alternative approach consists in calculating primary production using modelled or measured daily
51 time series of incident irradiance, sea ice transmittance and in-water vertical attenuation coefficients,
52 combined with photosynthetic parameters determined from P vs. E curves measured with short (under
53 two hours) incubations of seawater samples inoculated with ^{14}C . The latter two methods require
54 that the vertical profile of the irradiance experienced by drifting phytoplankton be appropriately
55 determined, which is challenging due to surface heterogeneity. Traditionally, one or very few $E_d(z)$
56 profiles are measured at discrete locations under sea ice (e.g., Mundy *et al.* [7]). Such measurements,
57 however, do not capture the variability induced by sea ice features. In recent studies, to better document
58 the spatial variability of $E_d(z)$, radiometers were attached to either remotely operated vehicles (ROV)
59 [4] or a surface and under-ice trawl (SUIT), a net developed for deployment in ice-covered waters,
60 typically behind an icebreaker [6]. Both an ROV and a SUIT allow a better description of the light field
61 right under sea ice, which is more appropriate for determining average irradiance experienced by
62 drifting phytoplankton. Such under-ice measurements can then be combined with averaged K_d values
63 to propagate light at depth.

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

79 than what would be expected from the inherent optical properties (IOPs) of the water column. In
 80 the field, this situation is more difficult to identify compared to profiles showing subsurface maxima
 81 because the former measurements may appear to follow a single exponential decrease but would not
 82 produce a diffuse attenuation coefficient that adequately describes the water mass. So, two vertical
 83 light profiles measured a few meters apart under sea ice are often very different. More importantly,
 84 local measurements of light under heterogeneous sea ice do not provide an adequate description of
 85 the average light field as it would be seen by drifting phytoplankton cells at different depths. This
 86 makes estimations of primary production and the interpretation of biogeochemical data challenging in
 87 the presence of sea ice.

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

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

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

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

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

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

113 2. Material and methods

114 2.1. Study site and field campaign

115 The field campaign was part of the GreenEdge project (www.greenedgeproject.info) which was
 116 conducted on landfast ice southeast of the Qikiqtarjuaq Island in the Baffin Bay (67.4797N, 63.7895W).
 117 The field operations took place at an ice camp where the water depth was 360 m, from April 20 to
 118 July 27, 2016 (Supplementary Fig. A1 included in Appendix). During the sampling period, the study
 119 site experienced changes in the snow cover and landfast ice thickness of 0-49 cm and 106-149 cm,
 120 respectively.

121 2.2. *In situ* underwater light measurements

122 During the campaign, a total of 83 vertical light profiles were acquired using a factory-calibrated
123 ICE-Pro (an ice floe version of the C-OPS, or Compact-Optical Profiling System, from Biospherical
124 Instruments Inc.) equipped with both downward plane irradiance $E_d(z)$ (W m^{-2}) and upward radiance
125 $L_u(z)$ ($\text{W m}^{-2} \text{sr}^{-1}$) radiometers. The ICE-Pro system is a negatively buoyant instrument with a
126 cylindrical shape 10 inches in diameter and is not designed for free-fall casts (as opposed to its
127 open-water version). To perform the profiles, the frame was manually lowered into an auger hole
128 that had been cleaned of ice chunks. Once it was underneath the ice layer, fresh clean snow was
129 shovelled back in the hole to prevent the creation of a bright spot right on top of the sensors. Great
130 care was taken not to pollute the hole surroundings (footsteps, water and slush spillage from the auger
131 drilling, etc.). The operator then stepped back 50 m, while keeping the sensors right under the ice, to
132 avoid any human shadow on top of the profile. The frame was then lowered manually at a constant
133 descent rate of approximately 0.3 m s^{-1} . The above-surface atmospheric reference sensor was fixed on
134 a steady tripod standing on the floe approximately 2 m above the surface and above all neighbouring
135 ice camp features. Data processing and validation were performed using a protocol inspired by the
136 one proposed by Smith and Baker [11]. Measurements were made at 19 wavelengths: 380, 395, 412,
137 443, 465, 490, 510, 532, 555, 560, 589, 625, 665, 683, 694, 710, 765, 780 and 875 nm. For this study, E_d and
138 L_u spectra were interpolated linearly between 400 and 700 nm every 10 nm. In situ diffuse attenuation
139 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
140 m, . . . , 70–75 m, 75–80 m) starting at 10 m depth to reduce the effects of surface heterogeneity. A total
141 of 72 044 non-linear models were calculated to estimate both K coefficients from Equation 1 (83 profiles
142 \times 14 depths \times 31 wavelengths \times 2 radiometric quantities (E_d, L_u)). A conservative R^2 of 0.99 was
143 used essentially to filter out noisy profiles. 42 407 models were kept for subsequent analysis.

144 2.3. 3D Monte Carlo numerical simulations of radiative transfer

145 2.3.1. Theory and geometry

146 3D numerical Monte Carlo simulation is a convenient approach for modelling the light field
147 under spatially heterogeneous sea surfaces [5,12–14]. They are simple to understand and versatile, and
148 incident light, IOPs and geometry can be easily changed. In this study, we used SimulO, a 3D Monte
149 Carlo software program that simulates the propagation of light in optical instruments or in ocean
150 waters [15]. Our objective was to simulate the propagation of sunlight underneath heterogeneous
151 ice-covered ocean waters. Simulations were performed in an idealized ocean described by a cylinder
152 of 120 m radius and 150 m depth (Fig. 1). The water IOPs were selected to reflect pre-bloom conditions
153 in the green/blue spectral region ($a = b = 0.05 \text{ m}^{-1}$). These typical averaged values were measured
154 during the GreenEdge 2016 campaign using an in situ spectrophotometer (ac-s from Sea-Bird Scientific)
155 and represent the contribution of both pure water and the water constituents. The scattering phase
156 function was described by a Fournier-Forand analytic form with a 3% backscatter fraction [16,17]. The
157 inclusion of a 3D sea ice layer at the upper boundary of the ocean would require extensive computing
158 power because of the high scattering properties of sea ice. Instead, sea ice was incorporated at the
159 upper boundary of the ocean using a 2D light-emitting surface with a radius of 100 m. The angular
160 distribution and magnitude of the light field emitted by the surface was chosen to mimic observed
161 field data [18]. SimulO does not allow the use of arbitrary angular distribution for photon-emitting
162 surfaces. To overcome this problem, two sources of photons were summed up in order to reproduce
163 an observed under-ice light field (Fig. 2). The first source was a regular Lambertian emitting surface
164 while the second was a Lambertian emitting surface but restricted to an emission within 60 degrees of
165 the zenith angle. A 5-m radius melt pond was set up at the center of the emitting surface (Fig. 1). The
166 melt pond had the same emitting angular distribution as the surrounding ice. Its intensity was four

167 times higher than the surrounding ice, which corresponds to typical conditions found in the Arctic
168 during summer [19].

169 Given our interest in surface light profiles, 2D horizontal software detectors were placed vertically
170 every 0.5 m, from 0.5 m up to a depth of 25 m. Detectors include 1 m² pixels measuring downward
171 irradiance and upward radiance (5-degree half angle of acceptance). In order to avoid the effect of the
172 boundary (i.e., absorption by the side of the cylinder used to simulate the water column), data outside
173 a radius of 50 m were not used (see the green box in Fig. 1). A total number of 7.14×10^{10} photons were
174 simulated to obtain a sufficient number of upwelling photons. The simulation took approximately
175 6 000 hours distributed over 2 000 CPU cores. Because the geometry was symmetrical azimuthally,
176 irradiance and radiance were averaged over the azimuth in order to increase the signal-to-noise ratio.
177 Because of the low scattering coefficients used to reproduce in situ conditions observed during the
178 sampling campaign, radiance profiles were noisy because a small number of upward photons could be
179 captured. To address this issue, radiance profiles were smoothed using a Gaussian fit (Supplementary
180 Fig. A2).

181 2.3.2. Estimation of reference and local light profiles

182 To explore how the melt pond influences the averaged underwater irradiance and radiance
183 profiles (Fig. 1), data from the Monte Carlo simulation were averaged according to six different radii,
184 corresponding to varying melt pond spatial proportions. The simulated light profiles were averaged
185 within the following surface areas: (1) 10 m radius (25% melt pond cover), (2) 11.18 m radius (20%
186 melt pond cover), (3) 12.91 m radius (15% melt pond cover), (4) 15.81 m radius (10% melt pond cover),
187 (5) 22.36 m radius (5% melt pond cover) and (6) 50 m radius (1% melt pond cover). For each of
188 these six configurations, the corresponding averaged light profile, $\overline{E_d}(z)$, was subsequently viewed
189 as an adequate description of the average underwater light field. For the remainder of the text, these
190 averaged profiles are referred to as reference light profiles. Furthermore, 50 light profiles, evenly
191 spaced by 1 m from the melt pond center, were extracted to mimic local measurements of light and to
192 calculate associated diffuse attenuation coefficients.

193 2.4. Statistical analysis

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

195 3. Results

196 3.1. Comparing in situ downward irradiance (E_d) and upward radiance (L_u) measurements

197 An example showing in situ downward irradiance (E_d) profiles and upward radiance (L_u) profiles
198 at 16 visible wavelengths measured under ice is presented in Fig. 3. For the E_d profiles, subsurface
199 light maxima at a depth of around 10 m are clearly visible between 400 and 560 nm. These peaks are
200 not visible in the yellow/red region (580–700 nm). For the L_u profiles, no subsurface light maxima
201 were found at any wavelength. To have a closer look at the shape of both E_d and L_u profiles, data
202 below the 10 m depth were normalized to the value at 10 m (Fig. 4). Below 10 m and between 400 and
203 580 nm, both E_d and L_u profiles presented the same shape (i.e., yield the same rate of attenuation with
204 increasing depth). At longer wavelengths (≥ 600 nm), differences between the shapes of E_d and L_u
205 profiles increased. Irradiance and radiance diffuse attenuation coefficients (K_d and K_{L_u}) calculated
206 for the layers of a 5 m thickness are compared in Fig. 5 for all 83 profiles. In the blue/green/yellow
207 regions (400–580 nm), the determination coefficients between K_{L_u} and K_d varied between 0.98 at the
208 surface (10–15 m) and 0.64 at depth (75–80 m). For most of the surface layers, regression lines lined
209 up with the 1:1 lines. Slight deviations from the 1:1 lines started to appear below 60 m where K_d was
210 on average higher than K_{L_u} . The relationships including orange and red wavelengths are presented
211 in Supplementary Fig. A3. A linear regression analysis between all in situ normalized E_d and L_u
212 profiles showed that determination coefficients (R^2) range between 0.75 and 1 (Supplementary Fig.

213 A4). A sharp decrease and a high variability of calculated R^2 occurred beyond 575 nm. This suggests
214 a gradual decoupling between E_d and L_u profiles at longer wavelengths, likely due to the effect of
215 inelastic scattering (mostly Raman scattering).

216 3.2. 3D Monte Carlo numerical simulations

217 Fig. 6 shows cross-sections of the simulated downward irradiance and upward radiance. A key
218 difference for the upcoming discussion is that the simulated upward radiance was more homogeneous
219 compared to the simulated downward irradiance. Fig. 7 shows the reference irradiance, $E_d(z)$, and
220 reference radiance, $L_u(z)$, profiles. The highest irradiance and radiance occurred when the melt pond
221 occupied 25% of the sampling area, allowing for more light to propagate in the water column. None of
222 the $E_d(z)$ and $L_u(z)$ reference profiles showed subsurface light maxima. Fig. 8 shows the 50 simulated
223 local downward irradiance and upward radiance profiles evenly spaced by 1 m in the horizontal
224 distance from the melt pond center. Local downward irradiance profiles under the melt pond (0–5
225 m) showed a rapid decrease with increasing depth described by a monotonically exponential or
226 quasi-exponential decrease. Local simulated downward irradiance profiles just outside the melt pond
227 (5–10 m from the melt pond center) were characterized with subsurface light maxima occurring at a
228 depth of between approximately 5 and 10 m. Further away from the melt pond center, downward
229 irradiance profiles followed a monotonically exponential or quasi-exponential decrease. None of the
230 simulated upward radiance profiles presented subsurface light maxima (Fig. 8). From local simulated
231 irradiance and radiance profiles (Fig. 8), K_d and K_{Lu} were calculated by fitting Equation 1 between
232 the depths of 0 m and 25 m. Results are presented in Fig. 9. K_d varied between 0.065 and 0.157 m^{-1}
233 and K_{Lu} between 0.079 and 0.116 m^{-1} . These K_d and K_{Lu} were used to propagate light downward
234 from surface reference values $E_d(0^-)$. Fig. 10 shows the profiles resulting from this calculation. A
235 greater dispersion around the reference profiles (thick black lines in Fig. 10) occurred when using K_d
236 compared to the profiles generated with similarly derived K_{Lu} values. The relative differences between
237 the depth-integrated values of each local profile (coloured lines in Fig. 10) and the depth-integrated
238 values of the reference profiles (thick black lines in Fig. 10) were used to quantify the error of using
239 either K_d or K_{Lu} as a proxy to predict downward irradiance in the water column (Fig. 11). Below
240 the melt pond, K_d overestimated the total downward irradiance by up to 40% when the melt pond
241 occupied 1% of the surface area. In this region, the local K coefficients are inflated. In the transition
242 region, at a horizontal distance of 5 and 10 m from the center of the melt pond, where subsurface
243 maxima are observed, K_d underestimated the downward irradiance by up to 35% when the melt pond
244 occupied 25% of the surface area. Further away from the edge of the melt pond, the errors saturated to
245 maximum of -25%. The same behaviour is observed for K_{Lu} but with about two times less amplitude.
246 The mean relative errors were lower by approximately a factor of two when using K_{Lu} (-7%) compared
247 to K_d (-12%). Also, the prediction errors stabilized at a shorter horizontal distance from the center of
248 the melt pond when using K_{Lu} (≈ 10 m) compared with using K_d (≈ 20 m).

249 3.3. Inelastic scattering

250 Based on in situ data, our results have pointed out that K_{Lu} is not a good proxy for K_d at longer
251 wavelengths (Supplementary Figs. A3 and A4) because of the effect of Raman scattering. To validate
252 this hypothesis, we used the HydroLight (Sequoia Scientific, Inc.) radiative transfer numerical model
253 to calculate theoretical downward irradiance and upward radiance and their associated vertical
254 attenuation coefficients in an open water column in the presence of Raman scattering. The simulation
255 was parameterized using IOPs measured during the field campaign (detailed information can be found
256 in the supplementary section entitled Raman inelastic scattering included in Appendix). The simulation
257 was able to reproduce the observed decoupling between K_d and K_{Lu} observed at wavelengths ≥ 600
258 nm (Supplementary Fig. A5). These results are generally consistent with previous findings from
259 radiative transfer simulations, which demonstrated the depth and spectral dependencies of diffuse
260 attenuation coefficients as affected by Raman scattering [21,22].

4. Discussion

In the Arctic, melt pond coverage, lead coverage, and ice and snow thickness can vary greatly in both time and space [23,24]. 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 meters over the horizontal distance because 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 the L_u radiometer measures upwelling photons coming from deeper depth, which have likely 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 of radiative transfer, 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 depth 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 center 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 center 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 [3,4,25], remote sensing techniques, and drone imagery. In this study, we used a Monte Carlo

310 approach to model an idealized surface with a single melt pond (Fig. 1, Fig. 6). Fig. 11 shows that the
311 effect of a melt pond with diameter 5 m is minimized at a horizontal distance of approximately 20 m
312 or more. Therefore, when many melt ponds are characterizing an area, if one has to perform a single
313 profile, measuring an upward radiance profile under bare ice as far away as possible from any melt
314 pond would minimize the error in estimating the area-averaged downward irradiance profile using
315 K_{Lu} . Although not representative of a complex Arctic sea ice surface, our simple surface geometry
316 allowed to study the transition from a high to a low transmission sea ice. Further 3D Monte Carlo
317 work could include a more complex geometry of heterogeneous surfaces.

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Figures

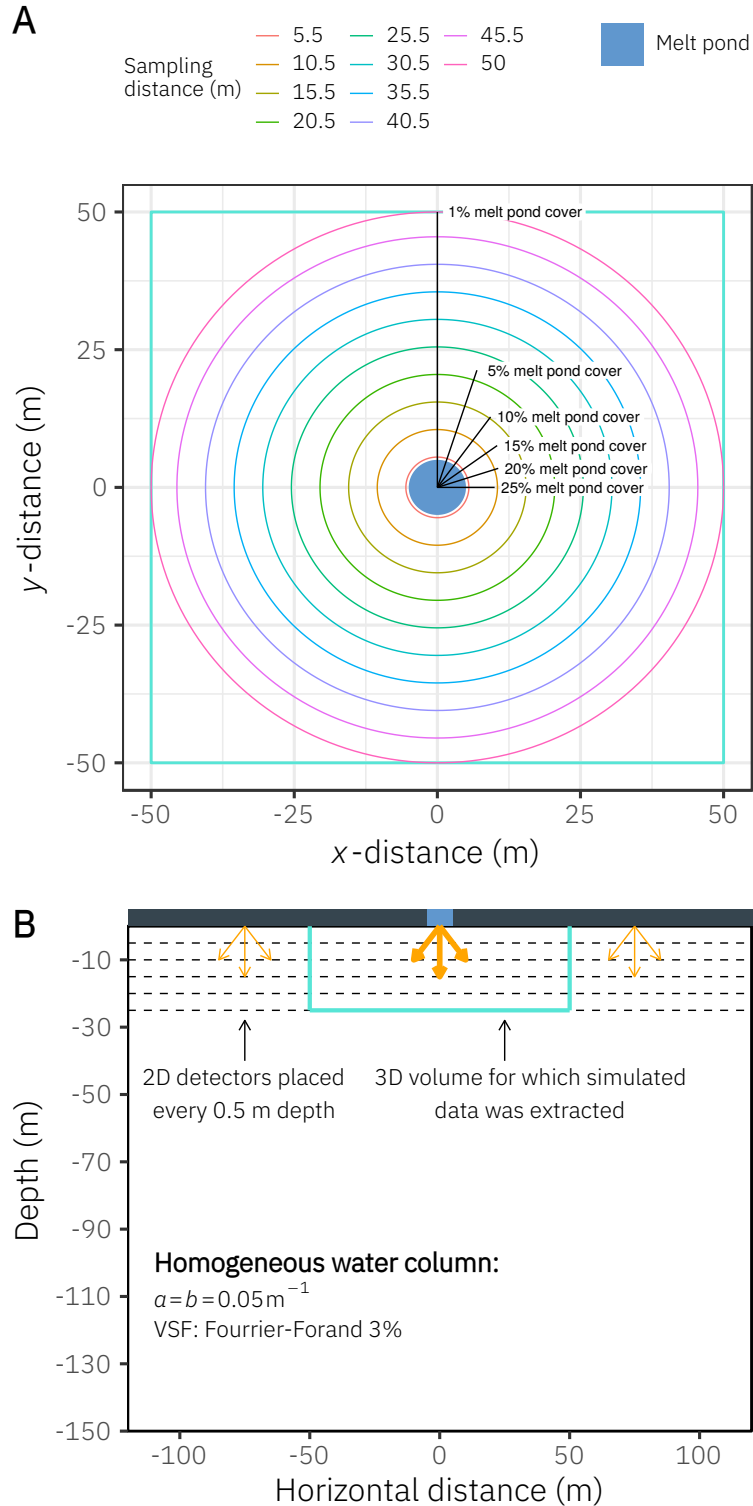


Figure 1: Spatial configuration used for the 3D Monte Carlo numerical simulations. **(A)** Surface view showing the percentage of the total area covered by the melt pond over the areas described by the black lines. For each of these areas, light profiles were averaged (see Fig. 7). For visualization purpose, lines of the horizontal sampling distances from the center of the melt pond have been plotted only at 5 m intervals. **(B)** 2D side view showing the 3D volume for which simulated data were extracted and how photon detectors were placed in the water column. Orange arrows indicate incident light sources.

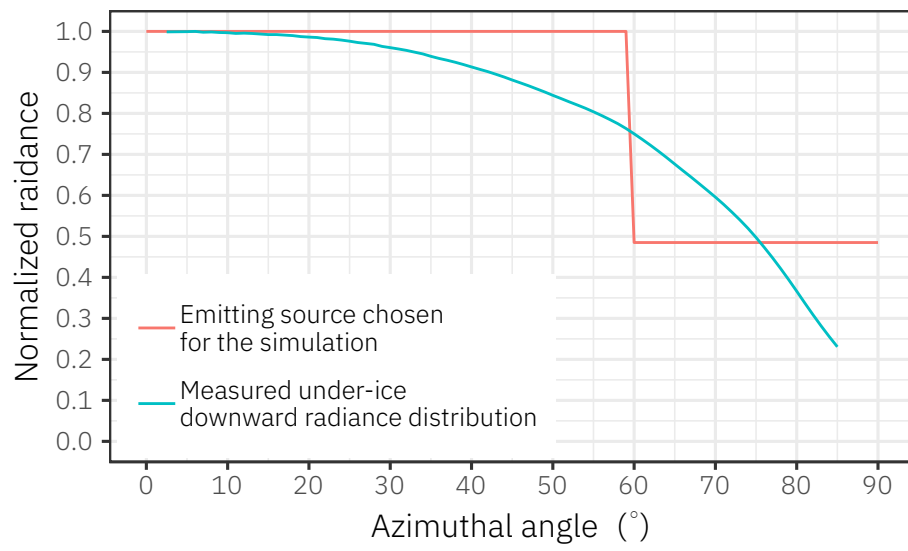


Figure 2: Comparison of the under-ice measured downward radiance distribution (the average cosine is ≈ 0.61 , Girard et al., 2018) and the angular distribution of light-emitting source used in the paper.

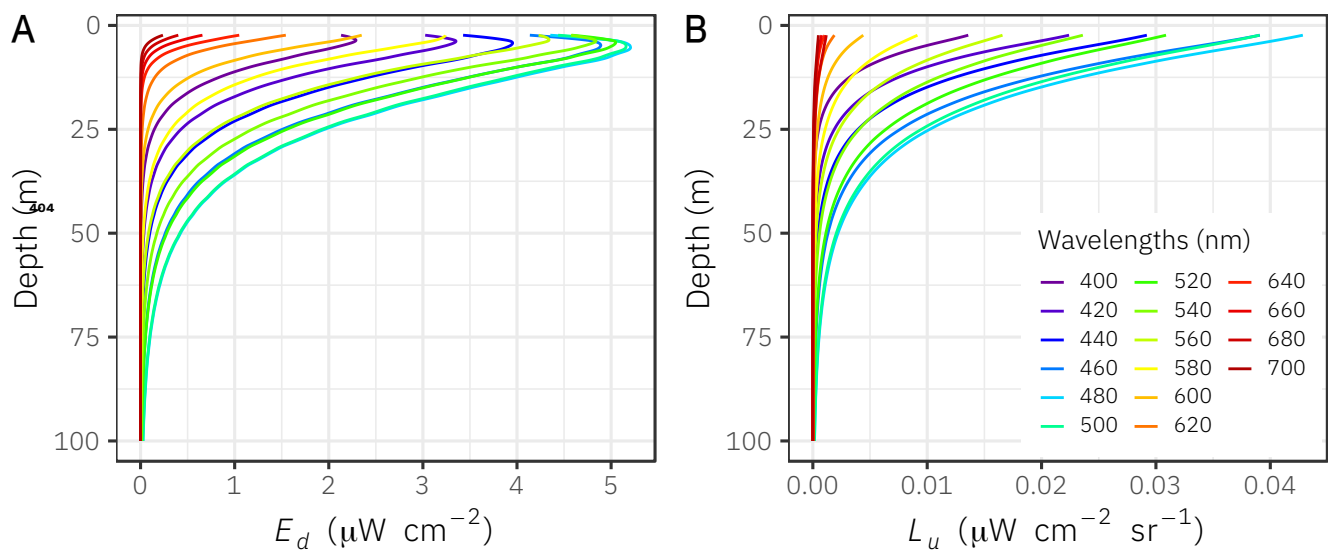


Figure 3: Examples of in situ downward irradiance ($E_d(z)$) and upward radiance ($L_u(z)$) profiles measured under-ice on 2016-06-20. Note the presence of subsurface maxima in the downward irradiance profiles and the absence of subsurface maxima in the upward radiance profiles.

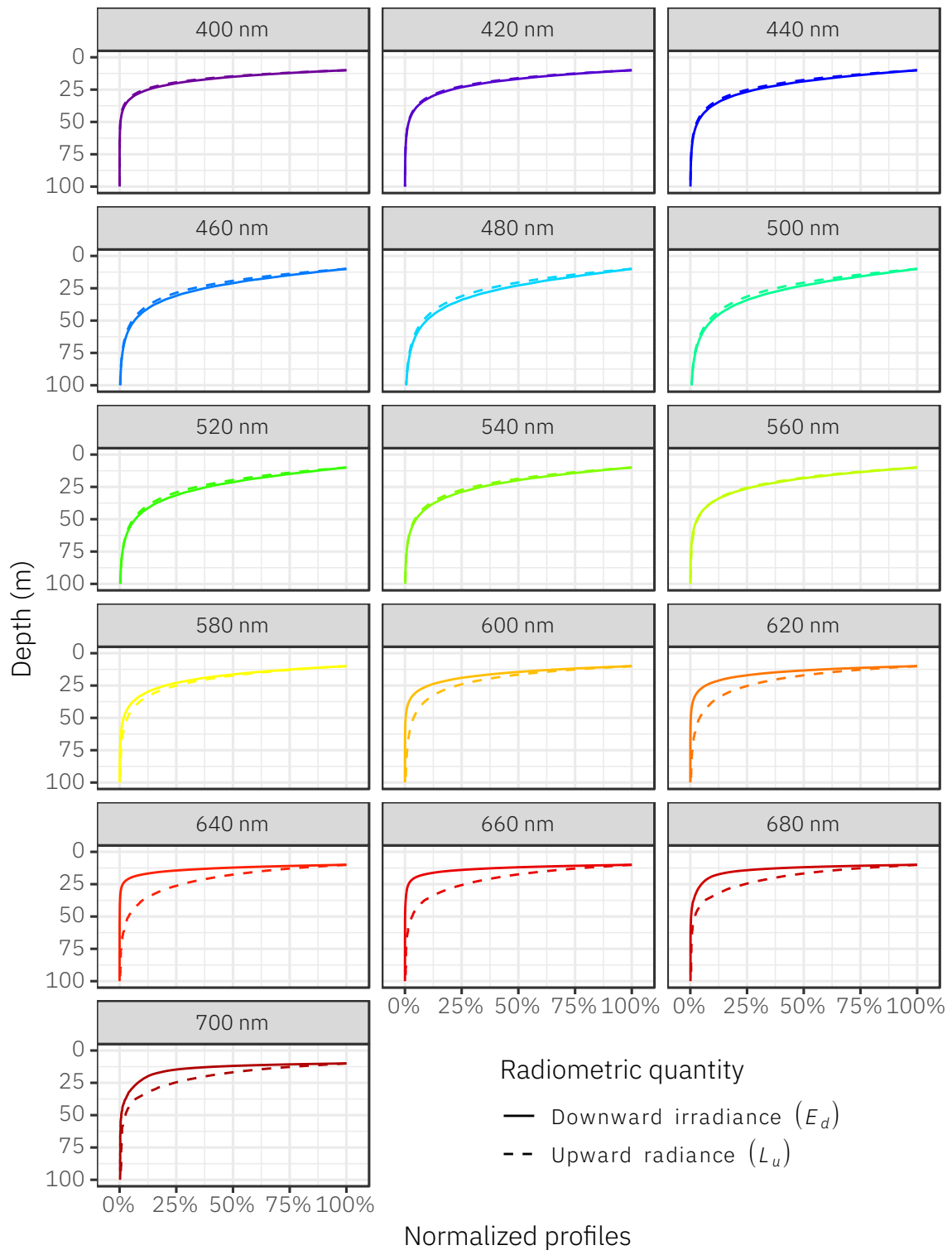


Figure 4: Comparison of downward irradiance ($E_d(z)$) and upward radiance ($L_u(z)$) for one example light profile measured under-ice. Profiles were normalized to the measured radiometric value at 10 m depth (under the subsurface light maximum) in order to emphasize the similar shape between $E_d(z)$ and $L_u(z)$.

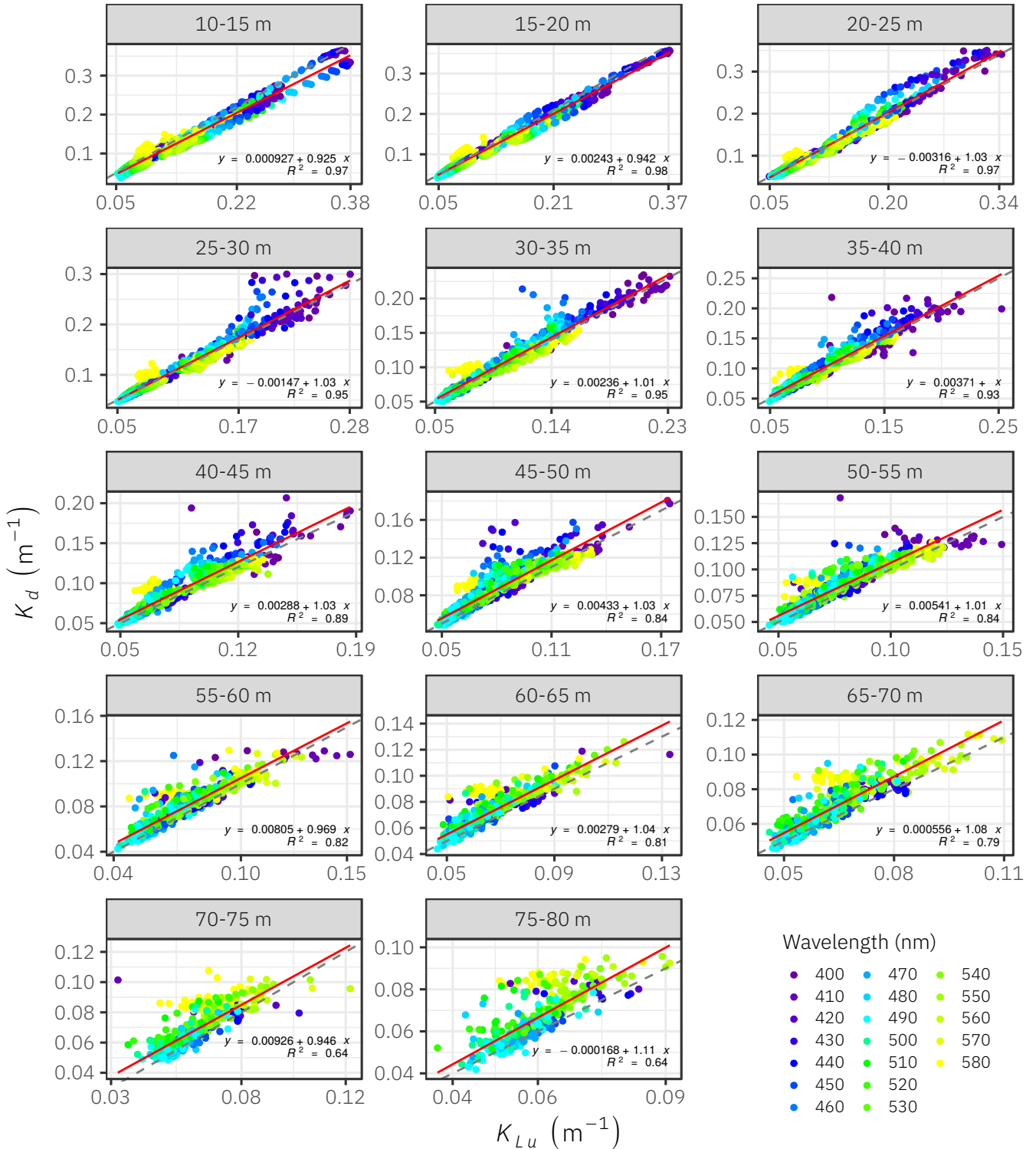


Figure 5: Scatter plots showing the relationships between the measured K_d and K_{Lu} in the spectral range between 400 and 580 nm at different depths (numbers in gray boxes). Red lines represent the regression lines of the fitted linear models. Regression equations and determination coefficients (R^2) are also provided in each plot. Dashed lines are the 1:1 lines.

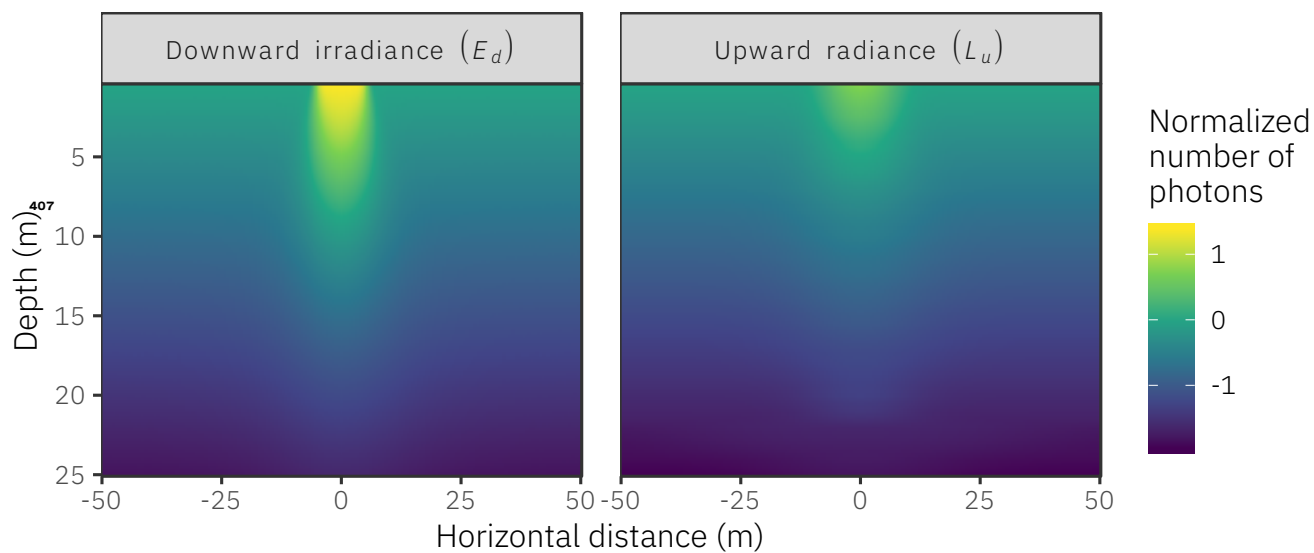


Figure 6: Cross-sections of simulated downward irradiance and upward radiance fields under a melt pond with a 5 m radius. The logarithm of the normalized number of photons has been used to create the scale for visualization. The normalization has been done using the values modelled at a 0.5 m depth and at a horizontal distance of 50 m from the center of the melt pond.

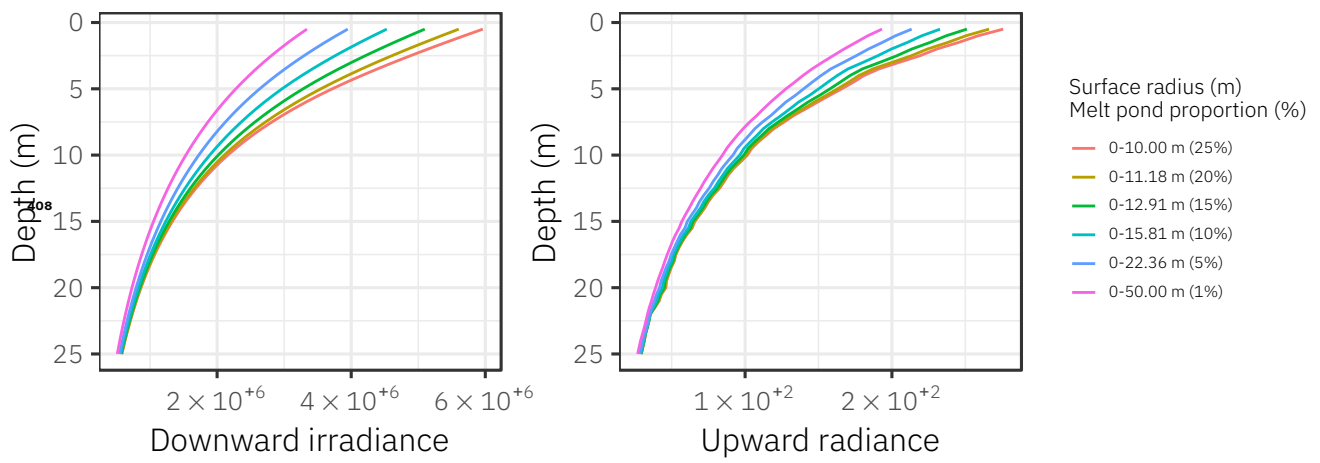


Figure 7: Simulated reference downward irradiance and upward radiance profiles ($\overline{E_d}(z)$, $\overline{L_u}(z)$ in relative units) for six different areas with varying proportions of the surface occupied by the melt pond (see Fig. 1). Note that none of the averaged irradiance profiles show the same subsurface light maxima as observed with in situ data (see Fig. 3).

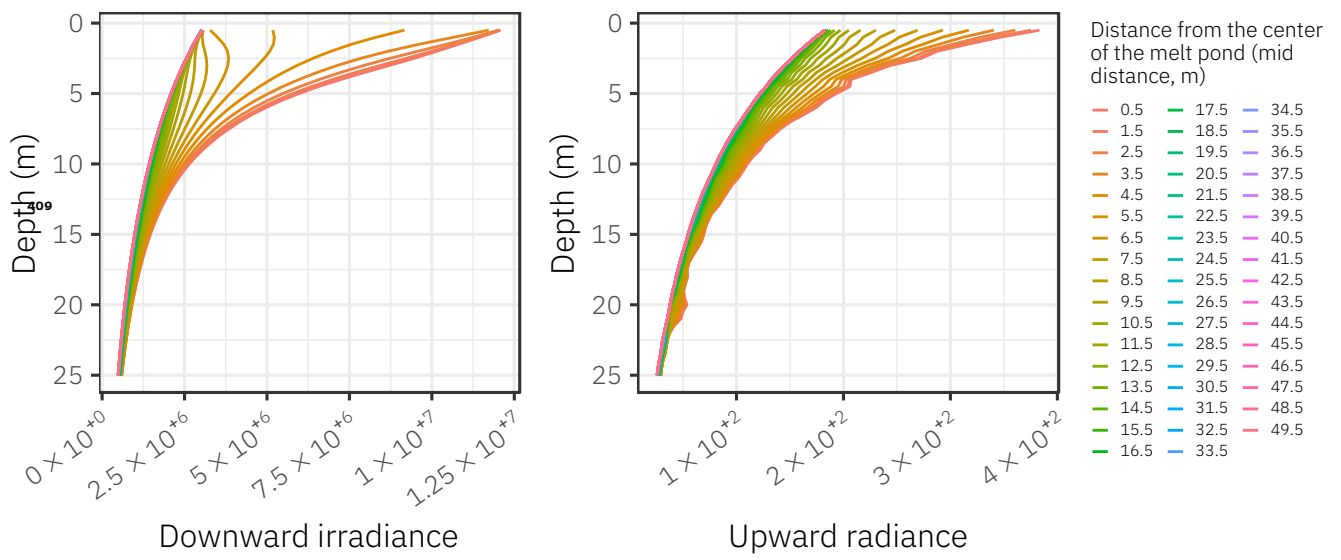


Figure 8: Simulated local downward irradiance and upward radiance profiles (expressed in relative units) at different horizontal distances from the center of the melt pond (see Fig. 1) used to compute K_d and K_{Lu} . These attenuation coefficients were used to propagate surface reference downward irradiance ($E_d(0^-)$), the surface values of the lines in Fig. 7) through the water column.

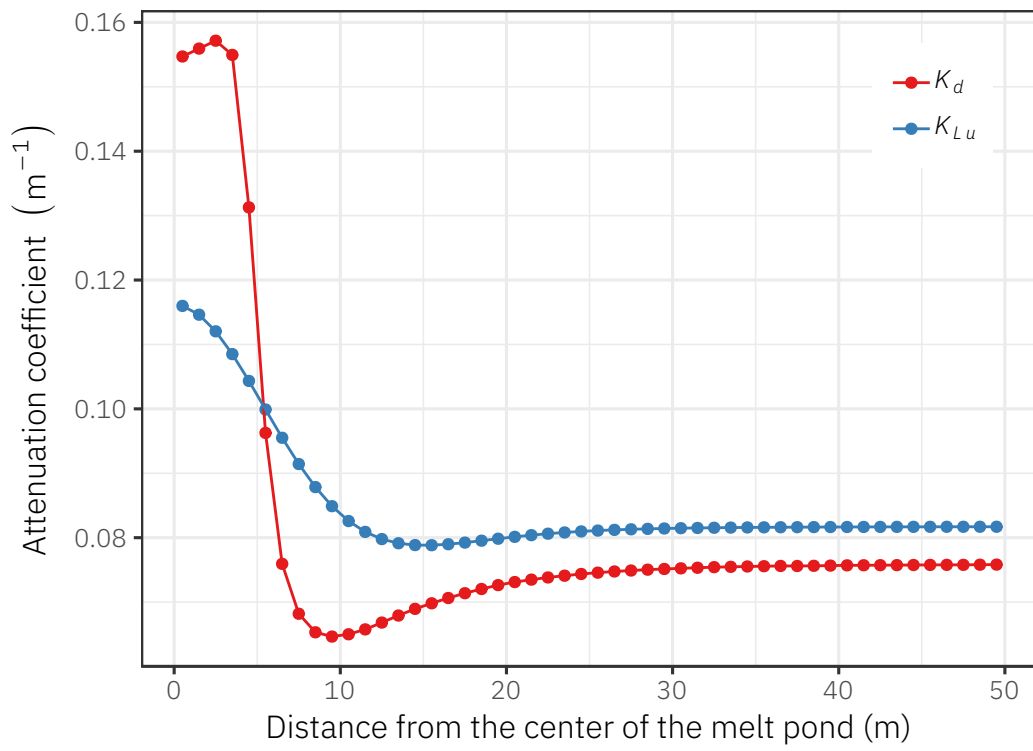


Figure 9: Diffuse attenuation coefficients calculated from local downward irradiance and upward radiance profiles simulated at different distances from the center of the melt pond (see Fig. 8).

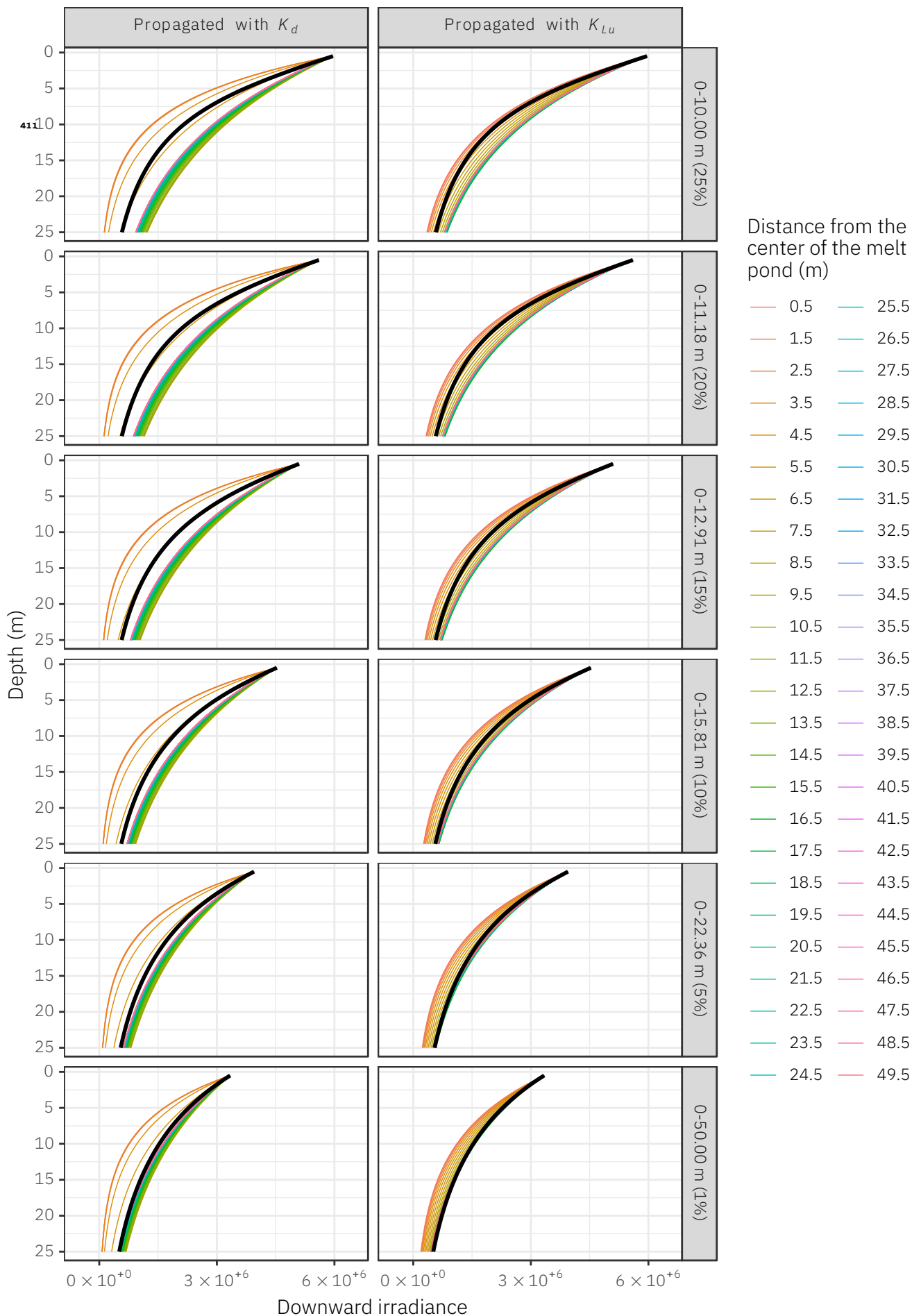


Figure 10: Reference downward irradiance profiles (thick black lines, in relative units) and propagated irradiance through the water column (colored lines, in relative units) using local values of K_d and K_{Lu} (see Fig. 8). Light was propagated using the surface reference downward irradiance.

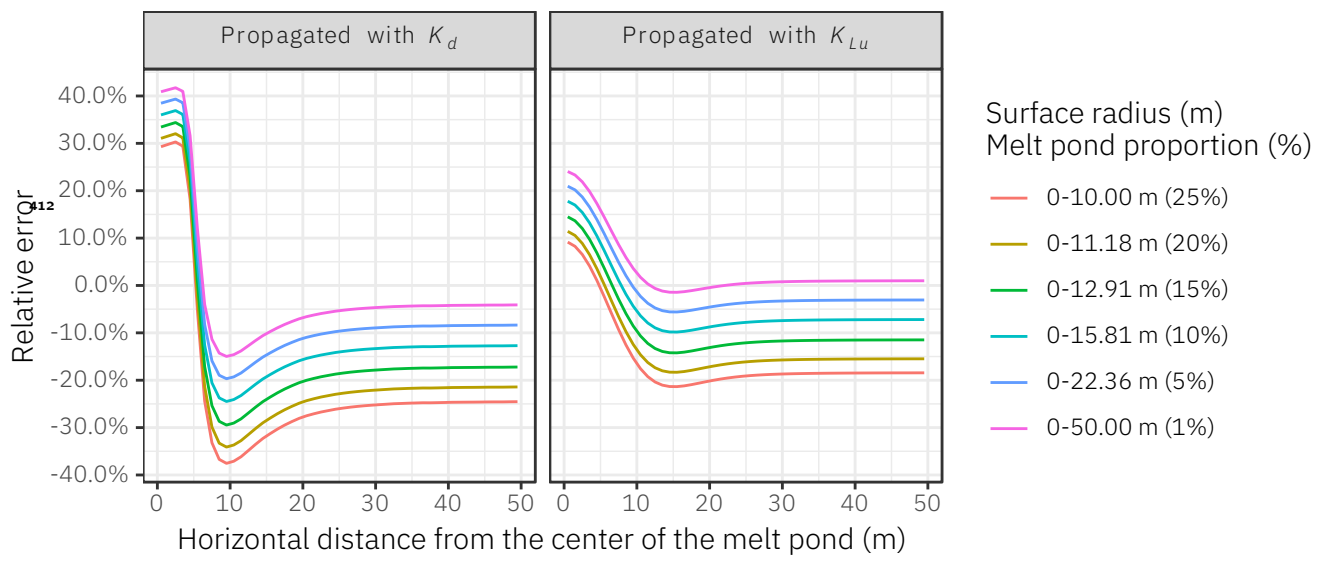


Figure 11: Relative errors of the predictions calculated as the relative differences between the depth integral of the reference and predicted irradiance profiles.

Appendix



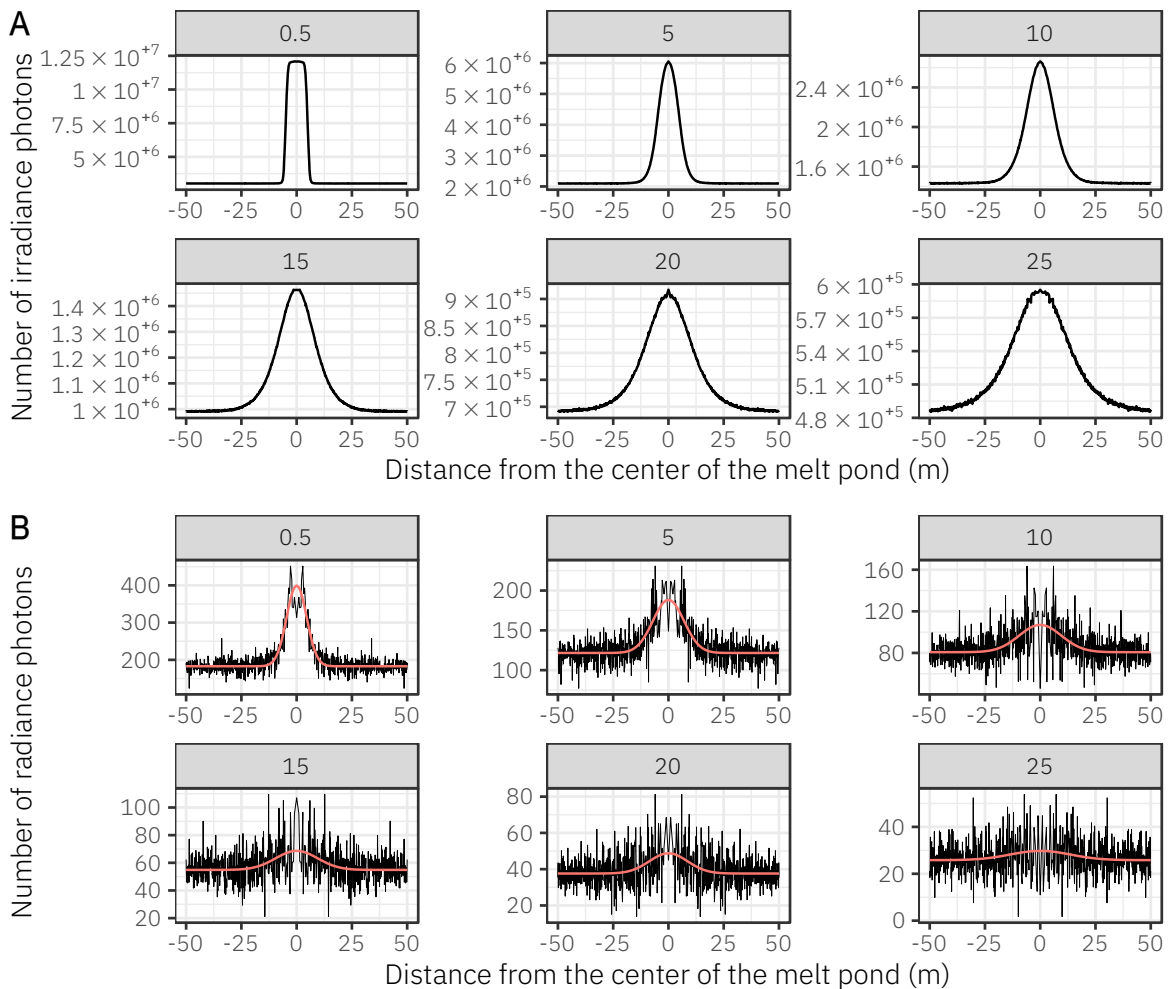
Supplementary Fig. A1: The field campaign was part of the GreenEdge project (www.greenedgeproject.info) which was conducted on landfast ice southeast of the Qikiqtarjuaq Island in the Baffin Bay (67.4797N, 63.7895W).

Smoothing radiance data

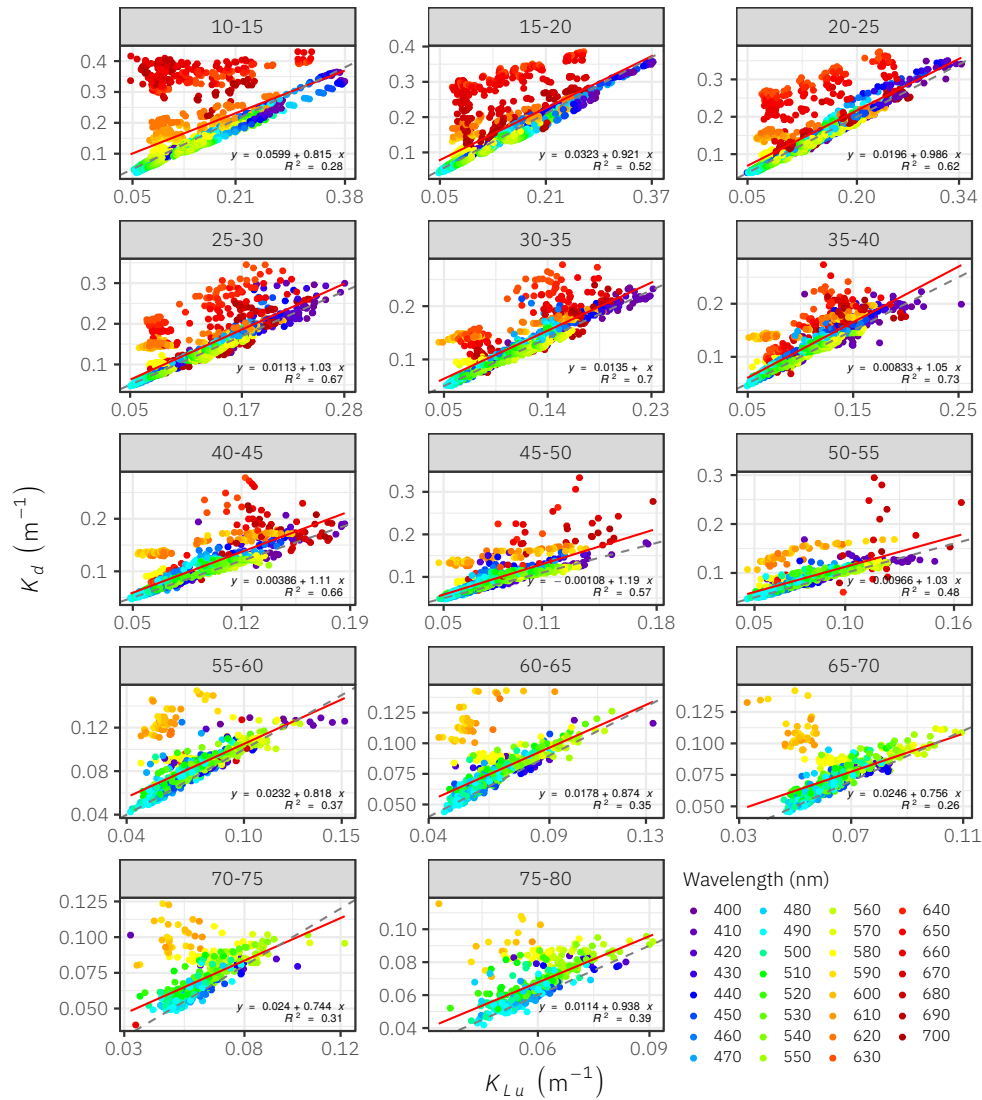
Due to the low scattering coefficients used to reproduce in situ conditions observed during the sampling campaign, radiance profiles were noisy because only few photons were scattered back in the upward direction (note the different y-scales). To overcome this problem, upward radiance data were smoothed using a Gaussian fit accordingly to Equation A1:

$$f(x, \varphi, \mu, \sigma, k) = \varphi e^{-\frac{(x - \mu)^2}{2\sigma^2}} + k \quad (\text{A1})$$

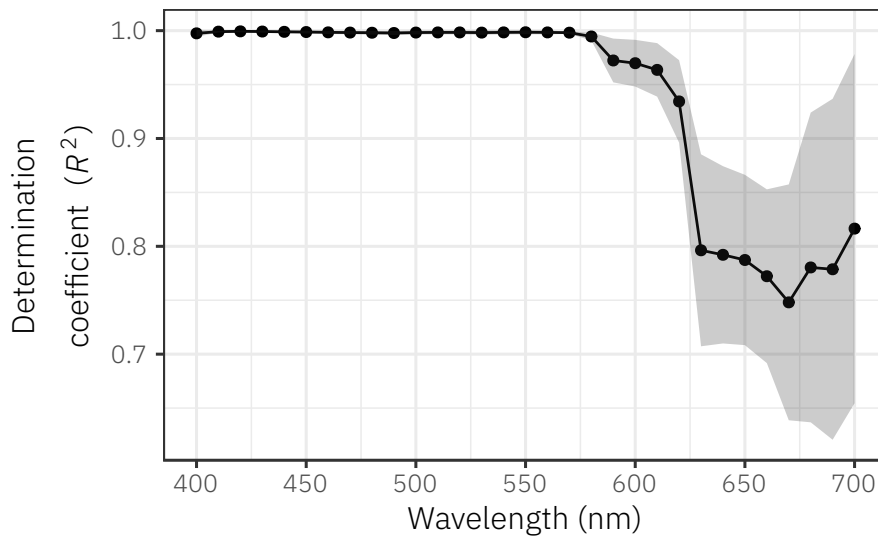
where x (m) is the horizontal distance from the center of the melt pond, σ (m) is the standard deviation controlling the width of the curve, φ is the height of the curve peak ($\varphi = \frac{1}{\sigma\sqrt{2\pi}}$), μ (m) is the position of the center of the peak, and k an offset coefficient.



Supplementary Fig. A2: Examples showing the number of downward irradiance (A) and upward radiance (B) photons captured by the detectors of the Monte Carlo simulation at different depth ranges (numbers in gray boxes) as a function of the horizontal distance from the melt pond. The red lines represent the fitted Gaussian curves.



Supplementary Fig. A3: Scatter plots showing the relationships between downward irradiance ($E_d(z)$) and upward radiance ($L_u(z)$) between 400 and 700 nm at different depths (numbers in gray boxes). Red lines represent the regression lines of the fitted linear models. Dashed lines are the 1:1 lines. Note the large deviations between the data points and the 1:1 line occurring in the orange and red regions (≥ 600 nm).



Supplementary Fig. A4: Average determination coefficient R^2 and standard deviation (shaded area) of the regressions between normalized (at 10 m depth) $E_d(z)$ and $L_u(z)$ profiles between 400 and 700 nm. At each wavelength, average values were computed from the 83 COPS measurements. A sharp decrease of R^2 occurred at wavelength longer than approximately 575 nm, suggesting a gradual decoupling between $E_d(z)$ and $L_u(z)$ profiles at longer wavelengths, possibly due to the effect of inelastic scattering.

Raman inelastic scattering

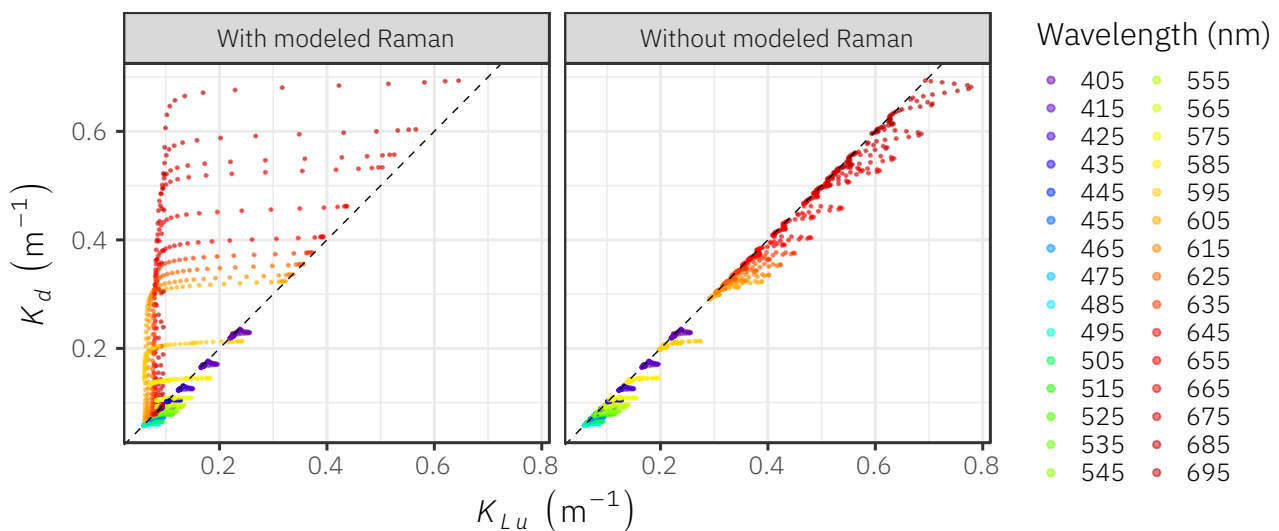
Raman scattering is a process by which photons, interacting with water molecules, lose or gain energy and are scattered at a different wavelength than the one they were originating from. In Supplementary Fig. A3 and Supplementary Fig. A4, one can observe a decoupling between K_d and K_{Lu} at longer wavelengths, possibly due to inelastic Raman scattering. To validate this hypothesis, we used the HydroLight radiative transfer numerical model to calculate downward irradiance and upward radiance and their associated attenuation coefficients in a water column.

HydroLight simulations

Two HydroLight simulations were carried out to model downward irradiance and upward radiance with and without taking into account Raman inelastic scattering. The simulations were parameterized using an IOPs profile (ac-s from Sea-Bird Scientific) measured on the first of May 2015 in the Baffin Bay. Simulations were performed with the following characteristics:

- A surface free of ice.
- A surface without waves.
- Sun position at noon for May 1st (solar zenith angle = 45.39 degrees).
- A cloudless sky.
- No fluorescence.
- Using HydroLight default atmospheric parameters.
- The scattering phase function of water was described by a Fournier-Forand analytic form with a 3% backscatter fraction.
- EcoLight option was run.

The HydroLight simulations showed a decoupling between K_d and K_{Lu} starting at around 600 nm when Raman scattering was modelled (Supplementary Fig. A5). Similar decoupling was also observed with the in situ data (see Supplementary Fig. A3).



Supplementary Fig. A5: Scatter plots showing the relationships between K_d and K_{Lu} calculated from the downward irradiance and upward radiance profiles modelled with and without Raman scattering. The dashed lines represent the 1:1 lines.

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