Airborne measurements of directional reflectivity over the Arctic marginal sea ice zone

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Abstract. The directional reflection of solar radiation by the Arctic Ocean is ~~dominated by two main~~ mainly shaped by two dominating surface types: sea ice (often snow-covered) and open ocean (ice-free~~(open) ocean . However, in the transitional~~). In the transitional zone between them, the marginal sea ice zone (MIZ), the ~~reflection properties~~ surface reflection properties are determined by a mixture of the reflectance of both surface types~~are mixed , which might cause~~ . Retrieval methods applied over the MIZ need to take into account the mixed directional reflectivity, otherwise uncertainties in the ~~results of retrieval methods of~~ retrieved atmospheric parameters over the MIZ ~~using airborne and satellite measurements~~may occur. To quantify these uncertainties, respective measurements of reflection properties of the MIZ are needed. Therefore, in this case study, an averaged hemispherical-directional reflectance factor (HDRF) of the inhomogeneous surface (mixture of sea ice and open ocean~~surfaces~~) in the MIZ is derived using airborne measurements collected with a digital fish-eye camera during a 20 minute low-level flight leg in cloud-free conditions. For this purpose, a sea ice mask was ~~constructed~~ developed to separate the reflectivity measurements from sea ice and open ocean ~~pixels. The separated data sets were accumulated and averaged to provide separate HDRFs for sea ice and open ocean surfaces~~and to derive separate HDRFs of the individual surface types. The respective results were compared with simulations and independent measurements available from the literature. ~~Using~~ It is shown that the open ocean HDRF in the MIZ differs from homogeneous ocean surfaces due to wave attenuation. Using individual HDRFs of both surface types and the sea ice fraction~~derived in parallel from the digital camera images~~, the mixed HDRF describing the directional reflectivity of the inhomogeneous surface of the MIZ was ~~reconstructed~~ retrieved by a linear weighting procedure. ~~The result was compared with the original measurements of directional reflectivity over the MIZ. It is concluded that the~~ Accounting for the wave attenuation, good agreement between the average measured HDRF and the constructed HDRF of the MIZ ~~can be well reconstructed by linear combination of the HDRFs of homogeneous sea ice and open ocean surfaces, accounting for the special conditions present in the MIZ compared to homogeneous surfaces.~~ was found for the presented case study.

1 Introduction

The Arctic Ocean is a key component of the complex Arctic climate system. From a solar radiative point of view, the surface of the Arctic Ocean is characterized by the strong contrast between the highly reflecting sea ice and the rather absorbing open ocean, which both determine the surface solar radiative energy budget. The reflection properties of both surface types are strongly linked to the sea ice–albedo-feedback, which significantly contributes to Arctic amplification (Pithan and Mauritsen, 2014; Wendisch et al., 2017). Often, the sea ice is covered by snow. The spectral albedo of the snow-covered Arctic sea ice typically varies between 0.8 and 0.9 in the visible spectral range (Wiscombe, 1980), depending on snow cover, grain size, melt pond fraction, and solar zenith angle (SZA). In contrast, solar radiation is mostly absorbed by open ocean. Its albedo depends on surface wind speed and SZA, and typically ranges below 0.1 for SZAs smaller than 65° in clear-sky conditions (Jin et al., 2004; Feng et al., 2016).

The reflection of solar radiation by snow-covered sea ice and open ocean surfaces was extensively studied and characterized by ground-based, airborne, and satellite observations (e. g., Cox and Munk, 1954; Gatebe et al., 2003; Bourgeois et al., 2006). However, large areas of the Arctic ocean are characterized by a mixture of sea ice and open ocean. The open water might be formed by leads or polynyas, while sea ice may also be covered by melt ponds (e. g., Hoffman et al., 2019). Also, in the marginal sea ice zone (MIZ), individual ice floes of different size cover varying fractions of the ocean. Strong and Rigor (2013) defined the MIZ as the zone where the sea ice concentration ranges between 0.15 and 0.8. In a warmer Arctic, the sea ice cover will further decrease (Perovich et al., 2018) and first-year sea ice will become more dominant. Accordingly, a significant reduction of the sea ice thickness was observed (Kwok, 2018). In summer, the thinner sea ice is more dynamic and breaks more easily, which leads to a higher amount of leads, a stronger drift, a faster dispersion of the floes, and a more extended MIZ (Kashiwase et al., 2017). Strong and Rigor (2013) showed that the width of the MIZ increased by about 13 km per decade in summer resulting in a widening of 39 % between 1979 and 2011. Accordingly, the summertime MIZ nowadays covers between 20 % and 60 % of the entire Arctic sea ice extent (Rolph et al., 2020). In winter, when the MIZ is dominated by freshly frozen sea ice, no significant trend of the MIZ extent was observed. The summertime MIZ widening trend highlights the necessity for characterizing the radiative properties of the mixture of sea ice and open ocean, which is needed to better quantify the complex radiative processes in such areas.

To quantify the solar radiative energy budget at the surface, the surface albedo as a hemispherically integrated measure of reflection is sufficient (e. g., Stapf et al., 2020). However, satellites detect the directional reflection instead of the surface albedo, such that conversion methods quantifying the surface reflection characteristics of all directions need to be applied (e. g., Schaaf et al., 2002). The directional distribution of the reflected radiances (as fraction of the incident radiation) is often quantified by the bidirectional reflectance distribution function (BRDF, mostly used in models) or the directly measurable hemispherical-directional reflectance factor (HDRF; Nicodemus et al., 1977; Schaepman-

Strub et al., 2006). While the BRDF only considers direct illumination from one single direction, also diffuse illumination from the entire hemisphere is taken into account by the HDRF.

The HDRF of homogeneous snow/sea ice and open ocean surfaces has been derived from measurements and simulations in numerous studies (e. g., Bourgeois et al., 2006; Gatebe et al., 2003). The HDRF of snow is characterized by high reflectance, which is further enhanced in forward direction (i.e., at high reflection zenith angles around 0° relative azimuth). This forward peak increases relative to the reflectance in nadir direction with increasing solar zenith angle and wavelength (e. g., Warren et al., 1998; Aoki et al., 2000; Carlsen et al., 2020). ~~Additionally~~Similar to the albedo, the snow HDRF depends on snow morphology (i. e., snow grain size and shape), which is crucial for accurate simulations of the snow HDRF (Jafariserajehlou et al., 2021). ~~In contrast, the open ocean is characterized by low reflectance because of the strong absorption of solar radiation by the liquid water. However~~Despite the low reflectance of open ocean for most reflection directions, in the so called sunglint angular range, specular reflection can be several orders of magnitudes higher. The width and maximum intensity of the sunglint area depend on surface roughness and, thus, on the wind speed. The higher the wind speed, the rougher the surface and the broader the sunglint area but the lower its maximum intensity (e. g., Cox and Munk, 1954; Jackson and Alpers, 2010).

Quantitative measurements of the HDRF of snow and open ocean were obtained by ground-based, airborne, and satellites observations. Bourgeois et al. (2006) and Marks et al. (2015) retrieved the HDRF of snow using a ground-based goniospectrometer, which measures the spectral radiance reflected from a small surface area from different directions. Their measurements provide an angular resolution of about 15° with a spectral resolution ~~of about~~ in the range of 1 nm. Commercial digital cameras equipped with a 180° fish-eye lens ~~provide instantaneous reflectance~~ deliver instantaneous radiance observations as a function of reflection directions of the lower hemisphere with high angular resolution ($< 0.5°$). In both cases, the downward spectral irradiance measured by a spectral hemispherical radiometer is required for the derivation of the HDRF. Goyens et al. (2018) combined ground-based ~~goniometer~~ goniospectrometer and fish-eye camera observations to benefit from both the hyperspectral ~~resolution and the~~ and hyperangular resolution. Since ~~goniometers~~ goniospectrometers are restricted to observations over solid ground, mostly shipborne or airborne measurements are used to derive the HDRF of open ocean (Cox and Munk, 1954). Gatebe et al. (2005) retrieved the open ocean wind speed and BDRF using the Cloud Absorption Radiometer (CAR), which is a scanning radiometer with a field of view (FOV) of 190°. This wide FOV is achieved by scanning the lower hemisphere with a narrow FOV optic of about 1° using a scanning mirror.

Airborne observations ~~with~~ using a digital camera equipped with a wide-angle lens have been reported by Ehrlich et al. (2012). Although the spectral resolution of the camera is limited ~~(only~~ to the three visible channels red, green and blue~~are available), this method~~, this technique was applied to analyze the impact of surface winds on the HDRF of the ocean surface. For homogeneous snow-covered areas in the Antarctic, Carlsen et al. (2020) used airborne 180° fish-eye camera observations to quantify the anisotropy of the snow HDRF with changing surface roughness, snow grain size, and SZA.

Most of the previous ground-based and airborne BRDF studies ~~published so far focused on a~~ focused on specific homogeneous surface ~~type~~types. The reflectivity of an inhomogeneous ~~mixture of different surface~~types~~surface~~surface, such as ~~present in~~ the MIZ, has rarely been observed~~in the past. Qu et al. (2016) linearly~~. Qu et al. (2016) combined simulated BRDFs of ~~the~~ different surface types in the MIZ in order to retrieve the surface albedo ~~in~~ of the MIZ from satellite observations. ~~This approach~~

is limited, since the open ocean surfaces are inhomogeneously distributed between the ice floes in the MIZ. Thus, However, ocean waves are attenuated (Kohout et al., 2011) and the dependence of the sunglint on the surface roughness and the wind speed could differ from ice-free open oceans. Therefore, in our

In this study, the spatial coverage of airborne observations with a 180° hemispherical coverage and high spatial resolution of airborne fish-eye camera is observations are used to characterize the HDRF of a mixture of open ocean and partly snow-covered sea ice surfaces in the MIZ. The observations have been obtained during the ACLOUD (Arctic CLoud Observations Using airborne measurements during polar Day) campaign northwest of Svalbard (Wendisch et al., 2019). The directional reflectance quantitiesAfter the introduction of the used quantities, as well as the observations are introduced in Sect. 2. In Sect. 3, the separation of the observed mixed scenes into the contributions of sea ice and open ocean is described by applying a sea ice mask. This allows the analysis and instruments in Section 2, the following two questions will be discussed: (i) How does the HDRF of the individual contributions of open ocean and sea ice in the MIZ differ from homogeneous surfaces? To answer this, the contributions of the individual HDRFs of sea ice and open ocean in the MIZ, which are presented in Sect. 4 together with a comparison surface types on the observed mixed scenes are separated applying a sea ice mask (Section 3) and compared to HDRFs of homogeneous surfaces . Finally, the separated HDRFs are recombined for different sea ice fractions (Section 4). (ii) Is it possible to combine the HDRF of the individual surfaces, weighted by the sea ice fraction, to obtain a representative MIZ HDRF? For this, the data set is divided into two subsets. The previously separated HDRFs of one subset are recombined and compared to the original observations in Sect. 5. (unseparated) mean of the other subset (Section 5).

2 Methodology and measurements

2.1 Definition of reflectance quantities

The spectral BRDF f_{BRDF} of a surface describes the directional distribution of the reflected radiation and is defined as:

$$f_{\mathrm{BRDF}}(\theta_{\mathrm{i}}, \phi_{\mathrm{i}}; \theta_{\mathrm{r}}, \phi_{\mathrm{r}}, \lambda) = \frac{\mathrm{d}I_{\mathrm{r}}(\theta_{\mathrm{i}}, \phi_{\mathrm{i}}; \theta_{\mathrm{r}}, \phi_{\mathrm{r}}, \lambda)}{\mathrm{d}F_{\mathrm{i}}(\theta_{\mathrm{i}}, \phi_{\mathrm{i}}, \lambda)} \tag{1}$$

(Nicodemus et al., 1977) . F_{i} represents the spectral irradiance (in $\mathrm{W\,m^{-2}\,nm^{-1}}$) illuminating (subscript "i") a surface at wavelength λ from the direction characterized by the incident zenith and azimuth angles, θ_{i} and ϕ_{i}, respectively. I_{r} quantifies the radiance (in $\mathrm{W\,m^{-2}\,nm^{-1}\,sr^{-1}}$) reflected (subscript "r") into the direction characterized by the reflection zenith and azimuth angles, θ_{r} and ϕ_{r}, respectively, and depends additionally on the incident angles. The BRDF has the unit of inverse steradiant ($\mathrm{sr^{-1}}$). Often the bidirectional reflectance factor (BRF) R_{BRF} is used instead of the BRDF. The reflectance factor (without unitdimensionless) is defined as the ratio of the BRDF of the actual surface to the constant BRDF of a Lambertian surface $f_{\mathrm{BRDF,id}}$, which is constantly equal to $1/\pi\,\mathrm{sr^{-1}}$. Thus:

$$R_{\mathrm{BRF}} = \frac{f_{\mathrm{BRDF}}}{f_{\mathrm{BRDF,id}}} = \pi\,\mathrm{sr} \cdot f_{\mathrm{BRDF}}. \tag{2}$$

Since the illumination under atmospheric conditions is a combination of a direct and a hemispherical diffuse irradiance component , F_{dir} and $F_{\mathrm{diff}}$with the fractions f_{dir} and $f_{\mathrm{diff}} = 1 - f_{\mathrm{dir}}$, respectively, both BRDF and BRF cannot be measured prac-

tically. Therefore, the hemispherical-directional reflectance factor (HDRF, ~~without unit~~dimensionless) R_{HDRF} is introduced (e. g., Schaepman-Strub et al., 2006)~~. However, if~~:

$$R_{\mathrm{HDRF}}(\theta_{\mathrm{i}}, \phi_{\mathrm{i}}, 2\pi; \theta_{\mathrm{r}}, \phi_{\mathrm{r}}) = R_{\mathrm{BRF}}(\theta_{\mathrm{i}}, \phi_{\mathrm{i}}; \theta_{\mathrm{r}}, \phi_{\mathrm{r}}) \cdot f_{\mathrm{dir}} + R(2\pi; \theta_{\mathrm{r}}, \phi_{\mathrm{r}}) \cdot f_{\mathrm{diff}}. \tag{3}$$

125 The reflectance factor of the diffuse radiation incident over the entire hemisphere is denoted by $R(2\pi; \theta_{\mathrm{r}}, \phi_{\mathrm{r}})$, where 2π refers to the diffuse radiation incidence. The direction of the direct component is given by θ_{i} and ϕ_{i}. The spectral dependence is omitted here. If the diffuse fraction of the incident radiation is sufficiently small, the HDRF represents a good approximation of the BRF. The ~~HDRF is determined by:~~

$$R_{\mathrm{HDRF}}(\theta_{\mathrm{i}}, \phi_{\mathrm{i}}, 2\pi; \theta_{\mathrm{r}}, \phi_{\mathrm{r}}) = \frac{\pi\,\mathrm{sr} \cdot I_{\mathrm{r}}(\theta_{\mathrm{i}}, \phi_{\mathrm{i}}, 2\pi; \theta_{\mathrm{r}}, \phi_{\mathrm{r}})}{F_{\mathrm{i}}(\theta_{\mathrm{i}}, \phi_{\mathrm{i}}, 2\pi)}.$$

130 infinitesimal quantities in Eqs. 1–3 are not measurable, in practice, measurement optics with sufficiently small opening angles are applied to approximate the finite radiances. Thus, from a measurement perspective, the HDRF is obtained by:

$$R_{\mathrm{HDRF}}(\theta_{\mathrm{i}}, \phi_{\mathrm{i}}, 2\pi; \theta_{\mathrm{r}}, \phi_{\mathrm{r}}) \frac{\pi\,\mathrm{sr} \cdot I_{\mathrm{r}}(\theta_{\mathrm{i}}, \phi_{\mathrm{i}}, 2\pi; \theta_{\mathrm{r}}, \phi_{\mathrm{r}})}{F_{\mathrm{i}}(\theta_{\mathrm{i}}, \phi_{\mathrm{i}}, 2\pi)}. \tag{4}$$

~~Here, the additional argument 2π refers to the diffuse radiation incidence over the whole hemisphere while θ_{i} and ϕ_{i} refer to the direction of the direct component. The spectral dependence is omitted here.~~

135 **2.2 Observations and instrumentation**

2.2.1 Airborne campaign

In this paper, data collected during the Arctic CLoud Observations Using airborne measurements during polar Day (ACLOUD; Wendisch et al., 2019) campaign, which took place in May and June 2017, are analyzed. During ACLOUD, the MIZ was located at about 80° N in the region north-west of Svalbard~~(Norway)~~. 19 measurement flights were conducted with each of
140 the two research aircraft *Polar 5* and *Polar 6* from Alfred Wegener Institute, Helmholtz Centre for Polar and Marine Research (AWI; Wesche et al., 2016). Both aircraft were equipped with downward-looking 180° fish-eye cameras measuring the upward directional radiances (Ehrlich et al., 2019; Jäkel et al., 2019). Additionally, on *Polar 5* the Spectral Modular Airborne Radiation measurement SysTem (SMART; Wendisch et al., 2001) was operated to measure the spectral downward solar irradiance.

The data analyzed ~~here~~ for the present case study were obtained during the flight on 25 June 2017 (flight number 23) per-
145 formed under cloudless conditions. Fig. 1a shows a satellite image from the MODerate resolution Imaging Spectroradiometer (MODIS) instrument onboard the satellite Terra during the research flight. The surface was dominated by an inhomogeneous distribution of sea ice floes with diameters of up to several kilometers. The flight track of *Polar 5* is plotted in red in Fig. 1a. A 20-minute leg performed between 12:26:35 UTC and 12:46:47 UTC (highlighted in yellow in Fig. 1a) was selected for the analysis, which ensures stable environmental conditions. The solar zenith angle ranged between 57.7° and 58.0° with a mean
150 of 57.8°. The observations were conducted along a 30 km straight flight section, during which the flight altitude varied between 65 and 165 m.
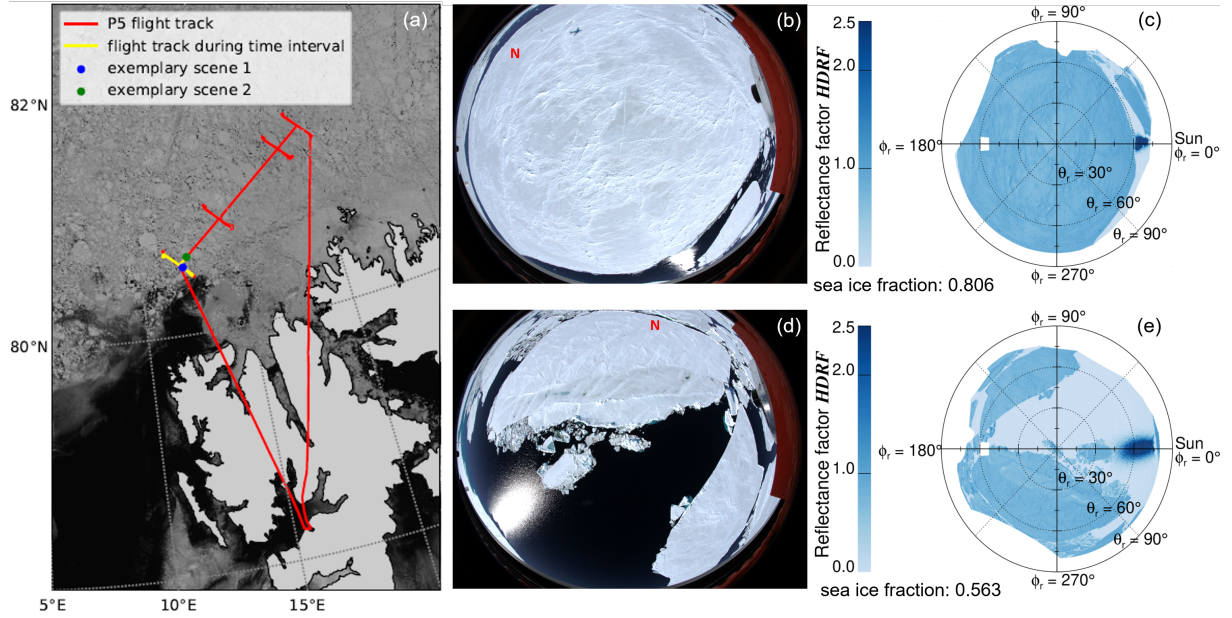
Figure 1. (a) Satellite image (composed of MODIS bands 1 and 2) of the MIZ north of Svalbard observed on 25 June 2017, 12:45 UTC including the flight track of *Polar 5* (red). The 20-minute period used for the analysis is highlighted in yellow. The blue and green dots ~~point at~~ indicate the locations where the exemplary scenes in (b) and (d), resp., were observed. (b) and (d) show exemplary images taken by the fish-eye camera at 12:37:23 and 12:46:41 UTC, resp. (c) and (e) present the processed HDRF polar plots of the scenes shown in (b) and (d), resp. Note that the polar plots are rotated compared to the raw images in a manner that the solar incident direction is always to the right of the plot.

2.2.2 180° Fish-eye camera

The upward radiance was measured by a CANON EOS-1D Mark III digital camera equipped with a 180° fish-eye lens. Images with a resolution of 3906×2600 pixels were taken every 6 s. As common for commercial cameras, each pixel covers three

155 spectral channels (RGB) centered at wavelengths of 591 nm (red), 530 nm (green), and 446 nm (blue) with a full width at half maximum (FWHM) of about 80 nm (Ehrlich et al., 2019; Carlsen et al., 2020). Figures 1b and 1d show two examples of raw true-color images taken by the fish-eye camera. The camera was calibrated in terms of geometrical, spectral and radiometric characteristics, which allows a conversion of the measured raw data into radiances as described in detail by Carlsen et al. (2020). In contrast to Carlsen et al. (2020), who applied a stellar method for the geometrical calibration, images of checkerboards

160 taken from different perspectives served as reference in this study. The images were analyzed by the open source routine `cv2.fisheye` from the free programming library OpenCV (http://opencv.org; Jiang, 2017). The backward model described by Urquhart et al. (2016) was applied to calculate the camera-fixed viewing zenith and azimuth angles, θ_v and ϕ_v, respectively, of each image pixel using the OpenCV output parameters.

The fish-eye camera was fixed to the aircraft frame. To obtain radiance measurements with respect to an Earth-fixed coordinate system, the viewing angles (θ_v, ϕ_v) of each camera pixel were corrected to consider the aircraft attitude angles (roll, pitch, yaw). Euler rotation matrices were applied to transform the viewing angles into the reflection angles (θ_r and ϕ_r) (Ehrlich et al., 2012). The azimuth plane of the images was rotated with respect to the relative position of the Sun, such that the Sun (and the forward direction) is located at the right in all polar plots shown in this paper. The footprint of one single image varied between 380 m and 915 m for the altitude range of the 20-minute leg assuming that the effective FOV of the fish-eye lens is 160°.

2.2.3 Calculation and uncertainty of the HDRF

Combining the downward irradiance F_i measured by SMART (Jäkel et al., 2019) and the angularly resolved radiances I_r from the fish-eye camera (Jäkel and Ehrlich, 2019) allows the calculation of the HDRF in flight altitude (Eq. 4), whereby the spectrally resolved irradiances were converted into the spectral range of each camera channel using the individual relative spectral response function from the spectral calibration.

The uncertainty of the radiometric calibration of the fish-eye camera was estimated by 4 % ~~(Carlsen et al., 2020). Further~~, further uncertainties stem from the sensor characteristics ~~, the geometric calibration, and the~~ and the combination of geometric calibration and aircraft attitude correction (0.5 % and 1 %, resp., Carlsen et al., 2020), leading to a total uncertainty of the fish-eye camera radiance measurements of about 4.2 %. However, during ACLOUD complementing radiance measurements were performed with SMART and a spectral imager, which demonstrated deviations of up to 35 % for the blue channel, while reflected radiances measured in the green and red channels ranged within the measurement uncertainties (Ehrlich et al., 2019). As a consequence, the fish-eye camera was inter-calibrated ~~. However, for the further analysis in this study only~~ with SMART. The radiances measured in the red channel ~~were selected, since they have shown~~ showed the best agreement ~~within the instrumental intercomparison . The root mean square deviation between the radiances of the digital camera and SMART~~ in the instrument intercomparison (root mean square deviation of $0.01\,\mathrm{W\,m^{-2}\,nm^{-1}\,sr-1}$ and the correlation coefficient of 0.98 between SMART and camera . The choice of the channel has only minor impact on the results presented here. While the reflectance of snow is spectrally neutral in the visible range, the spectral open ocean albedo is only slightly higher in the blue than in the red channel~~amounts to $0.01\,\mathrm{W\,m^{-2}\,nm^{-1}\,sr^{-1}}$ and the correlation coefficient is 0.98.~~. Therefore, it was decided to use the red camera channel in the following.

According to Bierwirth et al. (2009), the total uncertainty of the SMART irradiance measurements is 3.2 % in the visible spectral range. However, an updated transfer calibration (3 % error) and a larger cosine correction error (2 %) due to the larger solar zenith angle lead to an increased total uncertainty of 4.3 %. Thus, using the Gaussian error propagation, the total uncertainty of the calculated HDRF amounts to about 6 %.

Radiative transfer simulations performed with the library for radiative transfer ~~(libRadtran; Emde et al., 2016)~~ (libRadtran; Mayer and K were used to estimate the impact of the atmosphere between the ground and the maximum flight altitude (165 m) on the measured HDRF. They have shown, that the difference between the HDRF at ground and flight level is less than 1 % for the red camera channel.
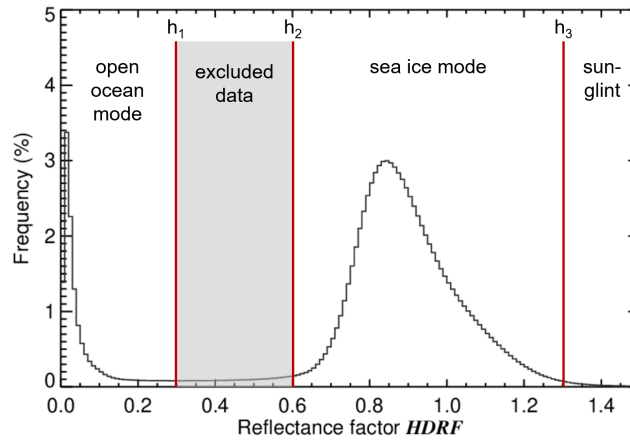
Figure 2. Frequency distribution of the observed HDRF for all directions and all images. The three vertical red lines indicate the thresholds of $h_1 = 0.3$, $h_2 = 0.6$ and $h_3 = 1.3$ applied in the sea ice mask algorithm.

The fully calibrated and processed HDRF obtained from the exemplary raw images in Figs. 1b and 1d are shown in Figs. 1c and 1e, respectively. The region contaminated by the shadow of the aircraft ($\theta_r \approx 60°$, $\phi_r \approx 180°$) was excluded and is represented by the white gap in the polar plots. The HDRF plots of the exemplary scenes reveal a high contrast between sea ice and open ocean areas and show a largely enhanced HDRF in the sunglint region of open ocean.

3 Separation of sea ice and open ocean surfaces

3.1 Sea ice mask

The set of images of a low-level flight section of 20 min was filtered for larger turns with roll or pitch angles larger than 5°. The remaining 138 fish-eye camera images ~~taken within the 20-minute period~~ were analyzed. Most of the individual images show a mixture of sea ice and open ocean. To obtain separate HDRFs for either surface type, a sea ice mask is constructed, which considers the different reflection characteristics of both surface types. Figure 2 (black line) shows a frequency distribution of all HDRF measurements, merging all images and all directions (pixels). The histogram reveals a distinct separation of the data into two modes, originating from open ocean and sea ice. Accordingly, all pixels with HDRF values below a threshold of $h_1 = 0.3$ are assigned to open ocean, pixels with HDRF values above a second threshold of $h_2 = 0.6$ are mostly related to sea ice but can also be assigned to open ocean in case of specular reflection observations. HDRF values between both thresholds, which are mostly linked to the ice floe edges, amount to roughly 3 % of the data and were excluded from further analysis.

In the sunglint region, also the HDRF of open ocean can feature values similar to or even exceeding typical values of sea ice. This sunglint area was identified by two additional criteria. Firstly, HDRF values larger than a third threshold of $h_3 = 1.3$ were considered as sunglint and assigned to open ocean. Secondly, a color ratio defined by the ratio of the radiances measured in the red (I_{red}) and the blue (I_{blue}) camera channel was used to identify the edges of the sunglint zone. Since the sunglint
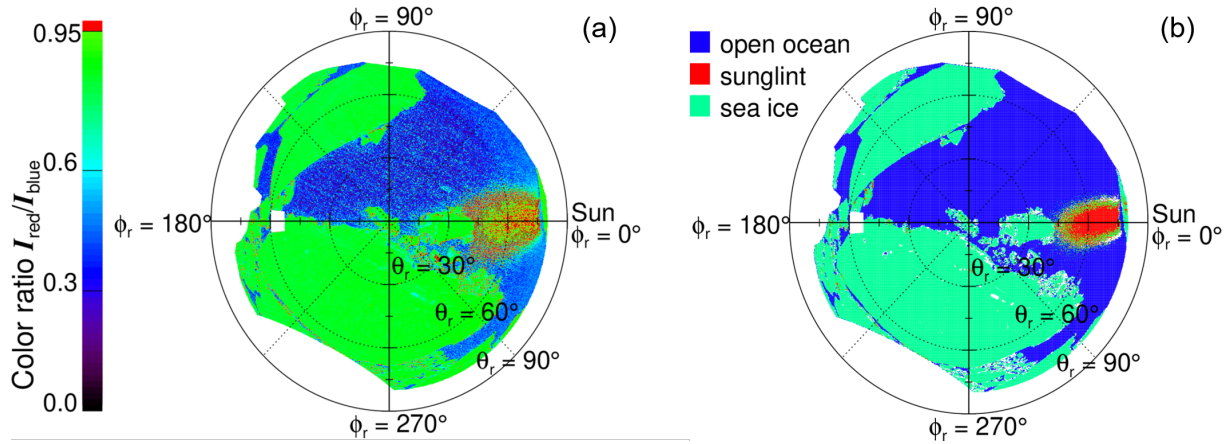
Figure 3. (a) Color ratio of an exemplary scene (same scene as in Fig. 1d). The area identified as sunglint (above the color ratio threshold $c = 0.95$) is highlighted in red. (b) Allocation of the pixels to sea ice (turquoise), open ocean (blue) or the sunglint area of open ocean (red) applying the sea ice mask illustrated in Fig. 4.

area appears more yellowish than the sea ice, its color ratio is higher. A color ratio threshold of $c = 0.95$ $c = I_{\text{red}}/I_{\text{blue}} = 0.95$ was chosen for the separation of sea ice and sunglint when the HDRF was between h_2 and h_3. In Fig. 3a the color ratio of the exemplary scene (same as in Fig. 1d) is shown with values higher than 0.95 highlighted in red. ~~Together with Fig.~~ Figure 3b ~~, which shows~~ illustrates the surface types identified by the sea ice mask~~, it demonstrates that the mask is capable to separate sea ice, open ocean, and sunglint. The complete decision process~~. Together, both panels show the capability of simple approach to separate between the surface types. Misclassifications mainly affect pixels at the sunglint margin (misclassified as sea ice), which do not have significant implications on the discussion and interpretation in this study. The uncertainty of the sea ice ~~mask is summarized in Fig. 4. An uncertainty analysis~~ fraction due to the limitations of the sea ice mask ~~with respect to the applied thresholds is given~~ is analyzed in the following section~~based on the calculated sea ice fraction~~. The complete decision process of the sea ice mask is summarized in Fig. 4.

3.2 Sea ice fraction

Using the sea ice mask, the sea ice fraction was calculated for each image. The sea ice fraction ~~usually refers to a horizontal surface~~ refers to the portion of the total horizontal surface that is covered by ice. The fish-eye lens of the camera, however, weighs the individual pixels equally although each pixel refers to a different surface area. Nadir pixels cover a smaller area than pixels close to the horizon. These different pixel projections were taken into account when the sea ice fraction averaged over a whole image was calculated.

Within the analyzed time interval, the sea ice fraction varied between 0.35 and 1.0. ~~Fig. 5 shows the~~ The frequency distribution of the ~~observed sea ice fractions.~~ derived sea ice fraction is shown in black in Fig. 5. Images with higher sea ice fractions were more frequent than images dominated by open ocean. The dashed line indicates the mean sea ice fraction of ~~the area~~
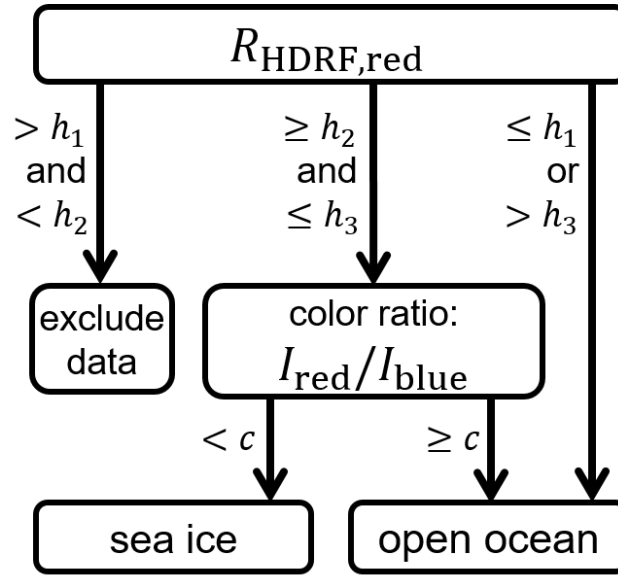
Figure 4. Decision tree separating pixels covered by sea ice and open ocean defining the sea ice mask. The applied thresholds $h_1 = 0.3$, $h_2 = 0.6$ and $h_3 = 1.3$ are based on the HDRF retrieved for the red camera channel $\cancel{HDRF_{\text{red}}}R_{\text{HDRF,red}}$. The threshold $c = 0.95$ is based on the ratio of the radiances measured in the red I_{red} and the blue I_{blue} channel.

235   0.83 sampled during the ~~20-minute~~ 20-minutes measurement time interval~~amounts to 0.83 (blue line in Fig. 5).~~. The accuracy of the sea ice fraction depends on the choice of the thresholds that are applied in the sea ice mask. In order to estimate the uncertainty related to the choice of the HDRF threshold values, a sensitivity study was performed, slightly varying one of the thresholds while the others were held constant. Two additional distributions, representing the lowest and the highest resulting sea ice fraction, are illustrated in Fig. 5.

240   The thresholds h_1 and h_2 were varied between the two modes (0.2 to 0.6). Changing h_1 or h_2 by 0.1 leads to a change in sea ice fraction of about 1.2 %. h_3 ~~was decreased down to 1.1. The~~ and the color ratio threshold c ~~was varied between~~ were varied between 1.2 and 1.4, and 0.9 ~~(lower values clearly misclassify sea ice as sunglint)~~ and 1.0, ~~whereby the sensitivity of~~ respectively. The sensitivity to the sea ice fraction is ~~larger when~~ higher, when h_3 or c ~~is decreased. The~~ are decreased. Hiowever, the lower limits of both h_3 and c were chosen such that the amount of obvious misclassification of sea ice as

245   sunglint is limited. As visible in Fig.5, the averaged sea ice fraction resulting from the variation of the thresholds ranges between ~~0.72~~ 0.79 and 0.86. ~~Since the misclassification of pixels is obvious for extreme cases~~Thus, the uncertainty of the sea ice fraction due to the sea ice mask is estimated to be less than 4 %.

Although the derived sea ice fraction slightly exceeds the upper limit of the MIZ definition given by Strong and Rigor (2013), the observations origin from an area very close to the sea ice edge and are characterized by separated and irregularly distributed

250   ice floes typical for the MIZ.
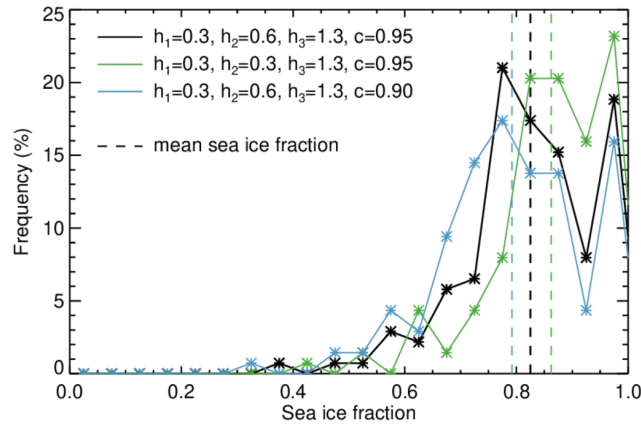
Figure 5. Frequency distribution of the sea ice fraction ~~resulting from~~ the applied sea ice mask for all images taken within the 20-minute time interval ~~, the~~ (black) and for adapted sea ice masks (color-coded). The vertical ~~blue line represents~~ dashed lines represent the resulting mean sea ice fraction~~of 0.83~~.

4 Separated HDRF and comparison with simulations and observations from the literature

After applying the sea ice mask to each of the images, all HDRF measurements of one single direction (pixel) assigned to either sea ice or open ocean were averaged. Doing so for all directions leads to a separated HDRF for each of the two surface types.

4.1 HDRF of open ocean

Figure 6a shows the average HDRF of the open ocean areas separated from the observations in the MIZ (separated open ocean HDRF). For most reflection directions the HDRF values are below 0.3. The average HDRF outside the sunglint region is 0.11 with an uncertainty of ± 0.02 when different thresholds are applied in the sea ice mask algorithm. In the sunglint region ($\theta_r \approx 60°$, $\phi_r \approx 0°$) the reflection is significantly enhanced and exceeds the maximum of the scale chosen here. A cross-section of the separated open ocean HDRF along the solar principle plane is illustrated in Fig. 6b (blue line) and shows the full dynamic range of the sunglint with a maximum HDRF of around 9. ~~Both the polar plot (Fig. 6a) and the cross-section (Fig. 6b) reveal a highly variable pattern of the entire separated open ocean HDRF.~~ The standard deviation (blue shading in Fig. 6b) is up to 0.6 outside the sunglint and up to 9.2 in the sunglint region. The high ~~variability is a result of the high sea ice fraction, which leads to a low number of open ocean~~pixels used for averaging. The sea ice fraction for each direction~~(pixel) of the solar principal plane~~ standard deviation outside the sunglint results from misclassifications of sea ice as "sunglint" which are, thus, assigned to open ocean. For each direction, the fraction of open ocean pixels classified as sunglint (calculated as ratio of all images) is referred to as pixel-based ~~sea ice fraction in the following and is~~ sunglint fraction and indicated by the ~~black~~ grey line in Fig. 6b. ~~Although there is also some directional variability in the~~ Even on the shadow side, about 10% of the pixels are contributions of misclassified sunglint. Neglecting the sunglint criteria (h_3 and c) would prevent from these misclassifications and reduce the
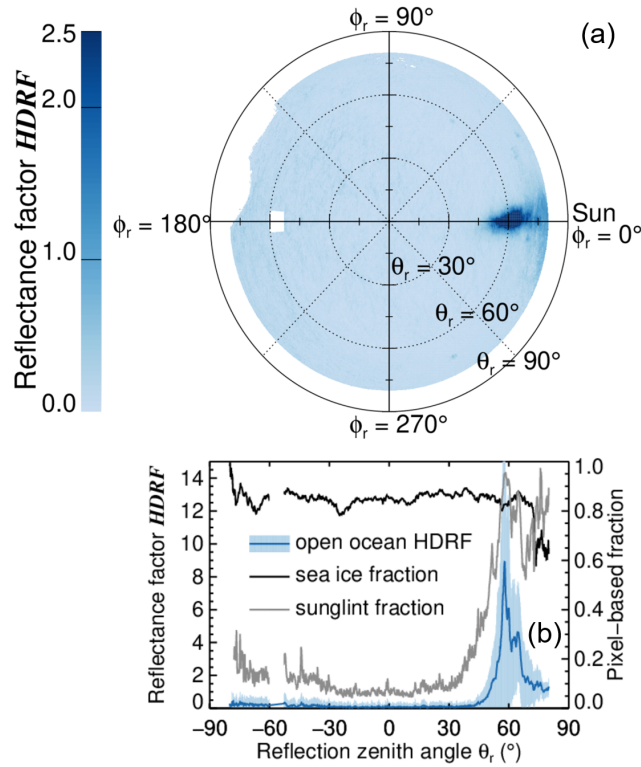
Figure 6. (a) Polar plot of the separated mean HDRF of open ocean. (b) Cross-section of the HDRF of open ocean along the solar principal plane (blue). The shaded area shows the standard deviation calculated for each direction of the solar principal plane from the entire time series (138 images). The black ~~and grey lines denote~~ line denotes the mean pixel-based sea ice fraction ~~and~~ (the portion of images, where the respective pixel of the solar principle plane is assigned to sea ice). The grey line denotes the mean pixel-based sunglint fraction ~~, resp.~~ (~~second y-axis~~the ratio of the number of images, where the respective pixel of the solar principle plane is classified as sunglint to the number of images, where this pixel is assigned to open ocean).

270 standard deviation to 0.11. Additionally, some of the observed variability results from the high sea ice fraction, which leads to a low number of open ocean pixels in the entire data set. Although the pixel-based sea ice fraction ~~, it~~ (black line in Fig. 6b), which is derived similarly to the pixel-based sunglint fraction, is higher than 0.8 for most reflection directions~~.~~, it also reveals significant directional variability.

In the observations presented here, the sunglint does not have the shape of a typical Gaussian distribution as reported in
275 literature (e. g., Cox and Munk, 1954; Gatebe and King, 2016; Ehrlich et al., 2012). Instead, several smaller local maxima are obvious beside a global one and the sunglint is slurred towards the horizon. The irregular shape of the sunglint is likely a result of the low ~~amount~~ number of observations and is imprinted in the pixel-based sunglint fraction~~(grey line in Fig. 6b). The pixel-based sunglint fraction is defined as the ratio of the number of observations identified as sunglint to the number of all observations assigned to open ocean by the sea ice mask for the respective direction.~~. It implies that the shape of the sunglint
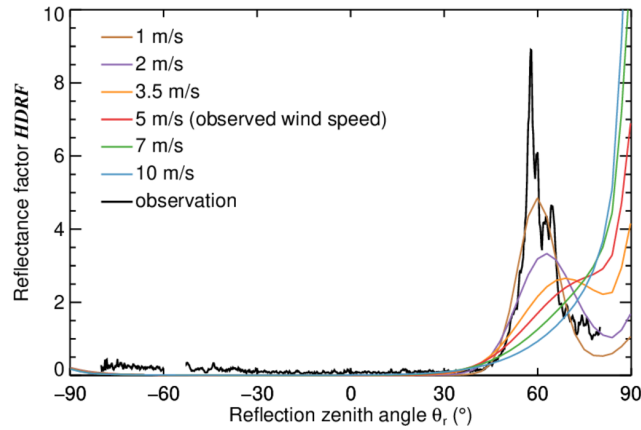
Figure 7. Cross-sections of the open ocean HDRF along the solar principle plane simulated with different wind speeds (color-coded) between 1 and $10\,\mathrm{m\,s^{-1}}$ including the observed wind speed of $5\,\mathrm{m\,s^{-1}}$. The black line corresponds to the separated open ocean HDRF. The mean observed SZA also used for the simulation was $57.8°$.

280 is highly variable among the images used for averaging, ranging from pure specular reflection of the Sun to sunglints cut by the edge of an ice floe (i. e., Fig. 1b) and widely blurred sunglints (i. e., Fig. 1d). The width of the sunglint primarily depends on the surface roughness, which is related to the surface wind speed (Cox and Munk, 1954). However, the distribution of sea ice and open ocean in the MIZ affects the surface roughness and its dependence on the wind speed. The development of waves is weaker in the gaps between ice floes than in the homogeneous open ocean (Kohout et al., 2011). This leads to the hypothesis

285 that the shape, extent and also position of the sunglint in the MIZ largely depend on the size of the open ocean areas between the ice floes. The ice floe distribution differed significantly between the images, which might have caused the diversity of the sunglints observed in these measurements.

The dependence of wind speed and sunglint was tested by comparing the separated open ocean HDRF to radiative transfer simulations. The simulated open ocean HDRF was obtained from libRadtran, which incorporates BRDF parametrizations

290 of different surface types (Mayer and Kylling, 2005) including open ocean (Cox and Munk, 1954). The simulations were performed for different wind speeds ranging from $1\,\mathrm{m\,s^{-1}}$ (minimum possible wind speed in libRadtran) to $10\,\mathrm{m\,s^{-1}}$. The atmospheric conditions were provided by measurements of a radiosonde launched at Ny-Ålesund at 12 UTC on 25 June 2017. Above, the default sub-Arctic summer atmosphere was selected. The solar zenith angle was set to $57.8°$, corresponding to the observation time. The hemispherical downward irradiance and the directional upward radiances at ground level (resolution

295 $3°$ in zenith and $5°$ in azimuth) were simulated in a wavelength range between 350 and $750\,\mathrm{nm}$ and were converted into the spectral range of the red camera channel.

Cross-sections of the simulated open ocean HDRF for several wind speeds are shown in Fig. 7. ~~Two~~ For low wind speeds, two local maxima of the open ocean HDRF are visible~~for low wind speeds. The first one~~, representing the sunglint (around the specular point~~represents the sunglint. It becomes broader and decreases in intensity as the wind speed increases. Furthermore,~~

300 ~~with~~) and the reflection of the diffuse incident radiation towards the horizon. With increasing wind speed, the ~~peak of the~~

~~sunglint~~ sunglint distribution becomes broader, while its maximum decreases in intensity and is shifted further ~~towards the horizon . This feature was already confirmed by observations from Su et al. (2002), although it is not present in the reflectances simulated by Jackson and Alpers (2010). The second maximum is characterized by an enhanced HDRF towards the horizon which~~ to the horizon (e. g., Su et al., 2002). The HDRF peak at the horizon increases with increasing wind speed. ~~This is likely~~
a ~~result of scattering in the atmosphere and multiple scattering at the sea surface. Radiative transfer simulations performed without atmospheric scattering (not shown here) indicate that the second maximum is reduced or even vanished (for low wind speeds, calm sea surface without multiple scattering),~~ which is likely due to an increase of diffuse incident radiation caused by multiple scattering between sea surface and atmosphere. For wind speeds higher than about $3.5\,\mathrm{m\,s^{-1}}$, the ~~second~~ diffuse reflection peak becomes dominant while the ~~first peak vanishes in the slopeof the second one.~~ sunglint vanishes in its slope.
The impact of the diffuse radiation (reflected skylight) on the BRDF and a method to remove this offset were discussed by Cox and Munk (1954). The black line in Fig. 7 represents the separated open ocean HDRF. The airborne measurements of the wind speed were extrapolated to $10\,\mathrm{m}$ over ground and amount to about $5\,\mathrm{m\,s^{-1}}$ on average. The comparison between the observation and the simulation for the observed wind speed shows that the maximum HDRF values and their positions do not coincide. Rather, the separated open ocean HDRF reveals a larger sunglint peak which is located closer to the specular point compared to the simulated open ocean HDRF. The HDRF simulated with a wind speed of $1\,\mathrm{m\,s^{-1}}$ (or even less) compares better to the observations. Unfortunately, simulations with lower wind speeds were not supported by libRadtran. This indicates that the actual surface roughness of open ocean in the MIZ is significantly reduced because of the wave attenuation between the ice floes, leading to a narrower and more intense sunglint with its maximum closer to the specular point. ~~Nonetheless, the maximum of the separated HDRF is still larger than that of the HDRF simulated with a wind speed of~~ The remaining discrepancy in maximum HDRF between observation and simulation ($1\,\mathrm{m\,s^{-1}}$. ~~The issue that the Cox-Munk simulations underestimate the sunglint intensity was also reported by Su et al. (2002) for large SZAs~~) could be due to the limited data set (e. g., the variability of the sunglint shape) or the still too high wind speed. However, ~~in contrast to their conclusion, a narrower simulated than observed sunglint was not found in our study, likely due to the variability of surface conditions within the period of observations.~~ since similar differences for high SZAs have been observed by Su et al. (2002), this might also be an effect of the larger SZA compared to the original measurements by Cox and Munk (1954).

Interestingly, the simulated open ocean HDRF outside the sunglint is significantly lower than the separated one. Nearly independent of the wind speed, the mean simulated HDRF of the shadow side ~~in the solar principle plane ($-90°$ to 0~~($90°$ to $270°$ reflection azimuth angle) is around ~~0.02~~0.03, which is about ~~one order of magnitude~~ 0.08 lower than for the separated HDRF. It is likely that these differences are due to horizontal photon transport. In the MIZ some photons reflected from the sea ice surface are scattered in the atmosphere such that they are detected in directions that actually point to open ocean as discussed by Schäfer et al. (2015). This effect may increase the HDRF of open ocean in the MIZ compared to homogeneous ice-free ocean.
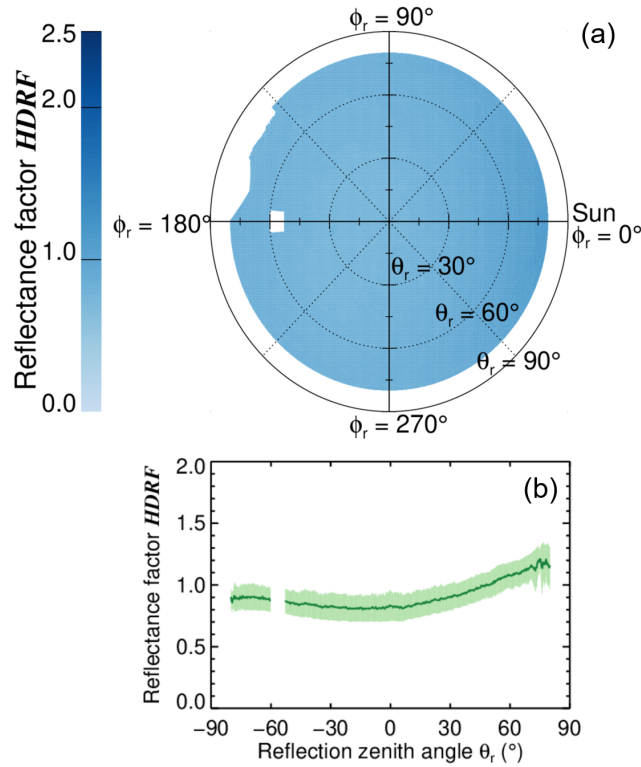
Figure 8. Same as Fig.6, but for sea ice ~~. The~~ and without the mean pixel-based sea ice fraction and the mean pixel-based sunglint fraction ~~are missing~~ in (b).

4.2 HDRF of ~~snow-covered~~ sea ice

Similar to open ocean (Fig. 6), the average HDRF of the separated snow-covered sea ice areas (separated sea ice HDRF) is shown in Fig. 8. Note, that the ranges of the y-axes in Figs. 6b and 8b are different. In comparison to open ocean, the HDRF values are significantly higher for a large part of the angular domain with values around 0.9. The HDRF is slightly enhanced in the forward direction, which is obvious in both the polar plot (Fig. 8a) and the HDRF along the solar principal plane (Fig. 8b), and is in accordance with the literature (e. g., Bourgeois et al., 2006; Gatebe et al., 2003; Goyens et al., 2018). ~~Furthermore, the variability of the separated sea ice HDRF (The maximum standard deviation below 0.20) is much lower than that of the separated open ocean HDRF due to the high sea ice fraction providing a higher number of images used for averaging. 0.2 reveals a lower variability compared to open ocean (0.6, including the misclassifications), which indicates that the mean sea ice HRDF is less affected by misclassified pixels.~~

In order to assess potential differences between the HDRF of homogeneous snow-covered sea ice surfaces and sea ice areas in the inhomogeneous MIZ, the separated sea ice HDRF is compared to homogeneous sea ice and snow HDRFs obtained from two studies (Goyens et al., 2018; Carlsen et al., 2020). The ground-based fish-eye camera measurements described by Goyens
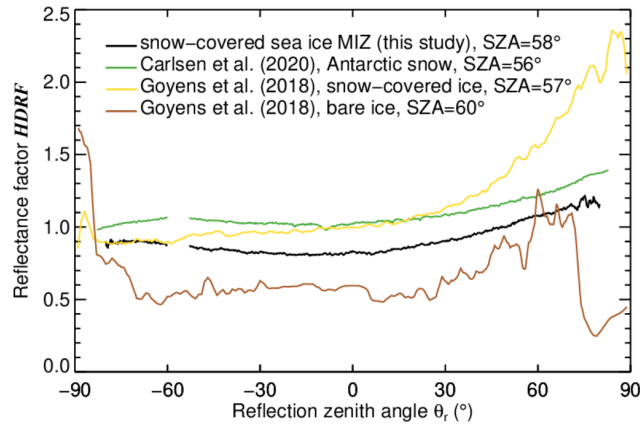
Figure 9. Comparison of the cross-sections of the HDRF of snow, snow-covered and bare sea ice along the solar principle plane obtained from different studies (color-coded), see text for details.

et al. (2018) were performed on landfast sea ice in the southern Baffin Bay in May and June 2015. The HDRF of three different sea ice surface types (bare ice, snow-covered ice and ponded ice) were analyzed. The airborne observations made by Carlsen et al. (2020) were performed over homogeneous snow surfaces on the Antarctic Plateau in December 2013. The SZA during these measurements was similar to the SZA observed in our case study (about 58°). The comparison of the HDRF along the
350   solar principle plane is shown in Fig. 9.

~~The snow HDRF from Carlsen et al. (2020) (green line) is larger than~~ Compared to the separated sea ice HDRF (black line), the snow HDRF fromCarlsen et al. (2020) has a similar shape, but shows a higher magnitude for all reflection directions of the solar principle plane ~~. Although the shape of both HDRFs is similar, the difference between them~~ (0.19 at nadir)~~decreases with increasing reflection zenith angle in the forward direction (min. 0.12 at 59°). The~~. Likewise, the HDRF of snow-covered
355   sea ice (yellow line) observed by Goyens et al. (2018) ~~agrees with~~ is larger than the separated sea ice HDRF ~~only~~ (0.16 at nadir), except for reflection zenith angles less than ~~−60°. With increasing reflection zenith angle, the difference between the HDRFsincreases to 0.95~~ −60°. However, comparing both snow HDRFs, significant differences in their anisotropies are obvious. While the ~~minimum~~ anisotropy of the snow HDRF measured by Carlsen et al. (2020) is lower than that of the separated sea ice HDRF ~~seems to be around nadir, the~~ (the difference between both reduces to 0.12 at 59°), the anisotropy is significantly larger
360   for the snow-covered sea ice HDRF ~~from Goyens et al. (2018) is lowest in the backward direction (at about −60°). However, the well documented increase of the HDRF towards the horizon in forward directionis also visible in their observations. The larger anisotropy~~ (Goyens et al., 2018) with a maximum difference to the separated sea ice HDRF of 0.95 at 77°. In contrast to the other HDRF distributions with a minimum in nadir viewing direction, the minimum of the snow-covered sea ice HDRF ~~observed by Goyens et al. (2018) can be explained by the smaller snow grain size (e. g., Warren et al., 1998). Although~~
365   ~~the exact optical-equivalent snow grain size is not known, the snow sampled~~ by Goyens et al. (2018) is ~~referred to as *cold snow* which typically has smaller grains than snow in the MIZ. Analyzing sections with high sea ice fraction, the snow grain size retrieval following the approach by Zege et al. (2011) showed grain sizes of about 200~~located in backward direction (at

about −60°). The reasons for the anisotropy differences of both snow HDRFs (Carlsen et al., 2020; Goyens et al., 2018) remain unclear and might result from, e.g., snow grain size, impurity load or surface roughness. While the measurements from Carlsen et al. (2020) were performed at a wavelength of 538 nm (green channel), the HDRF at 628 nm (red channel) by Goyens et al. (2018) were used for comparison. However, the spectral dependence of the snow HDRF in the spectral range is small. The increased variability of the snow-covered sea ice HDRF observed by Goyens et al. (2018) might be due to the smaller footprint of the ground-based measurements compared to the airborne observations. In particular, small-scale surface roughness features can be resolved, which contribute to the variability of the ground-based measurements. In contrast to the snow-covered surfaces, the HDRF of bare ice (brown line) is significantly lower than the separated HDRF of the airborne observations and is characterized by an increased anisotropy. This is most prominent at reflection zenith angles of about 60°. However, the shape of the bare ice HDRF distribution is less smooth and shows a variability, that is even larger than that of the snow-covered sea ice HDRF. According to Goyens et al. (2018), this is due to the presence of thawed ice nearby highly reflective ice grains, which often occurs at the beginning of the melt season.

The magnitude of the separated sea ice HDRF analyzed in this study ranges between the literature values for snow-covered and bare ice HDRF. This is reasonable since the observed ice floes revealed a mixture of snow-covered and bare ice (e.g., Fig. 1c). The comparison of the different HDRFs illustrates the variability of the snow and sea ice HDRF in polar environments, which is affected by a variety of properties (e.g., snow cover, snow grain size or impurity concentration). Due to the significantly larger area fraction of sea ice compared to open ocean, the impact of the nearby darker open ocean surfaces in terms of horizontal photon transport should be much smaller (< 0.02) for sea ice and can, thus, not completely explain the differences between the analyzed HDRFs.
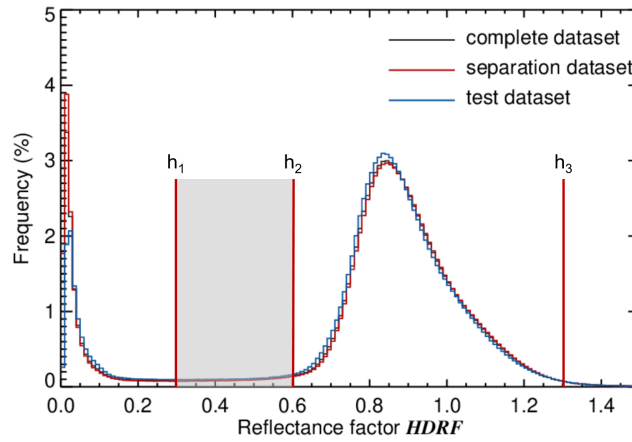
Figure 10. Same as Fig. 2, but also including the distributions of the separation data set (red) and test data set (blue).

5 HDRF of the MIZ as function of sea ice fraction

In ~~the next step~~this section, the average HDRF of the inhomogeneous sea ice–open ocean surface ~~of the~~in the MIZ is compared to a constructed HDRF of the MIZ assuming a linear combination of the individual HDRFs of open ocean $R_{\mathrm{HDRF,ocean}}$ and sea ice $R_{\mathrm{HDRF,ice}}$ weighted by the sea ice fraction f_{ice}:

$$R_{\mathrm{HDRF,recon}}(f_{\mathrm{ice}}) = f_{\mathrm{ice}} \cdot R_{\mathrm{HDRF,ice}} + (1 - f_{\mathrm{ice}}) \cdot R_{\mathrm{HDRF,ocean}}(f_{\mathrm{ice}}). \tag{5}$$

To do so, the data set was randomly split into two subsets. One of the subsets, the test data set, consists of 35 images (roughly 25 %) that are averaged without separation to obtain a mean HDRF of the inhomogeneous MIZ (mean MIZ HDRF~~, obtained by averaging all images without separation) is analyzed (~~). The sea ice fraction of this data set was calculated using the sea ice mask. The remaining subset, separation data set, was used to separate and recombine the individual HDRFs to a constructed HDRF. The HDRF histograms of both subsets are shown in Fig. 10. The location of the modes of both data sets is similar to the distribution of the entire data set (black line). The open ocean mode of the test data set is lower, while its sea ice mode is slightly shifted towards lower values. Nevertheless, the agreement of both data sets suggests, that the same thresholds of the sea ice mask can be applied to all images. The sea ice fraction amounts to 0.83 and 0.81 for the separation and the test data set, respectively, with uncertainties of 4 % similar to the complete data set.

The mean MIZ HDRF from the test data set is illustrated in Fig. 11a~~)~~. Despite the high sea ice fraction, the mean MIZ HDRF shows features of ~~the HDRFs of~~ both open ocean and sea ice surfaces. The strongly enhanced reflectance in the sunglint region is clearly visible. However, because of the high sea ice fraction, its maximum HDRF (about ~~2.6~~3.0) is significantly lower compared to the separated open ocean HDRF (see Fig. 6a). Outside the sunglint but still in forward direction, the slightly enhanced HDRF characteristic for the sea ice surface is imprinted in the mean MIZ HDRF (compare Fig. 8a). For all other directions, the HDRF is more or less isotropic with values slightly lower (mean of ~~0.74~~0.76 on the shadow side) than observed in the sea ice HDRF (~~0.85~~0.90), due to the contribution of open ocean surfaces.

~~In the following~~For comparison, the HDRF of the MIZ is ~~reconstructed (HDRF$_{recon}$) assuming a linear combination of individual HDRFs of open ocean HDRF$_{ocean}$ and sea ice HDRF$_{ice}$ weighted by the sea ice fraction f_{ice}:~~

$$HDRF_{recon}(f_{ice}) = f_{ice} \cdot HDRF_{ice} + (1 - f_{ice}) \cdot HDRF_{ocean}(f_{ice}).$$

constructed using Eq. 5. Firstly, it is tested if the mean MIZ HDRF ~~observed in this case study~~ from the test data set can be reproduced~~by the reconstructed MIZ HDRF calculated with Eq. 5 and using the observed sea ice fraction and the separated open~~

~~.~~ The MIZ HDRF is constructed using the separated open ocean and sea ice HDRFs ~~(Figs. 6 and 8). Then, the reconstructed MIZ HDRF is calculated for arbitrary sea ice fractions. To account for the dependence between surface wind speed, sea ice fraction and sunglint in the MIZ, simulations of HDRF$_{ocean}$ for different f_{ice} are used here. These simulations are performed in the same way as described in Sect. 4.1 except that the input surface wind speed v_{eff} is parametrized as a linear function of the sea ice fraction:~~

$$v_{eff} = v_{meas} \cdot (1 - f_{ice}).$$

~~v_{eff} is considered as an effective wind speed, that would produce the same surface roughness and, thus, the same open ocean HDRF if the ocean was ice-free. v_{meas} is the wind speed measured at flight altitude and scaled to 10 m. It has to be noted that this very basic relation between surface wind speed and~~ from the separation data set and the sea ice fraction ~~aims only to illustrate the effects in a qualitative view. Numbers may change if observations for different sea ice conditions are available.~~

~~The reconstructed MIZ HDRF calculated for $f_{ice} = 0.83$ (Eq. 5)~~ of the test data set (0.81). The constructed HDRF is shown in Fig. 11b and ~~is~~ compared to the mean MIZ HDRF (Fig. 11a). The difference between both HDRFs is less than 0.1 for ~~more than 96~~84 % of the pixels. ~~For more than 86 % of the pixels, the difference lies within the uncertainty range of the HDRF measurements (6 %). Polar plot of (a) the mean MIZ HDRF (obtained by averaging over the entire sequence of images without separation), and (b) the reconstructed MIZ HDRF (calculated from the separated HDRFs of sea ice and open ocean according to Eq. 5 using a sea ice fraction of 0.83). (c) Cross-section of the reconstructed MIZ HDRF along the solar principal plane obtained from the separated HDRFs for different sea ice fractions f_{ice} (color-coded) and the mean MIZ HDRF obtained without separation (black). (d) same as (c), but sea ice fraction-dependent simulations were used for the open ocean HDRF, see text for details.~~ ~~The reconstructed~~ The constructed MIZ HDRF appears more smoothed than the mean MIZ HDRF for statistical reasons. The smoothness was quantified by the standard deviation of the HDRF calculated with respect to all reflection directions of the shadow side (to exclude the sunglint contribution). For the ~~reconstructed~~ constructed MIZ HDRF the standard deviation is slightly lower (~~0.035, 4.7~~0.03, 4.3 % of the mean value) than for the mean MIZ HDRF (~~0.044, 5.9~~0.05, 6.5 % of the mean value). The ~~difference between both HDRFs originates from the calculation of the reconstructed MIZ HDRF with the mean~~ smoothness for the constructed HDRF is due to the assumption a homogeneous and uniform sea ice fraction ~~,~~ ~~where homogeneous conditions are assumed.~~~~This means that the~~ In Eq. 5, whereas the measured mean HDRF is affected by the directionally inhomogeneous sea ice fraction ~~is assumed to be uniform for each pixel (direction). In the observations this homogeneity is not given due to the limited~~ and the quite low number of images ~~used for averaging~~included in the test data set. Figure 12 ~~shows a polar plot~~illustrates the variability of the pixel-based sea ice fraction ~~, that was introduced in Sect. 4~~of the
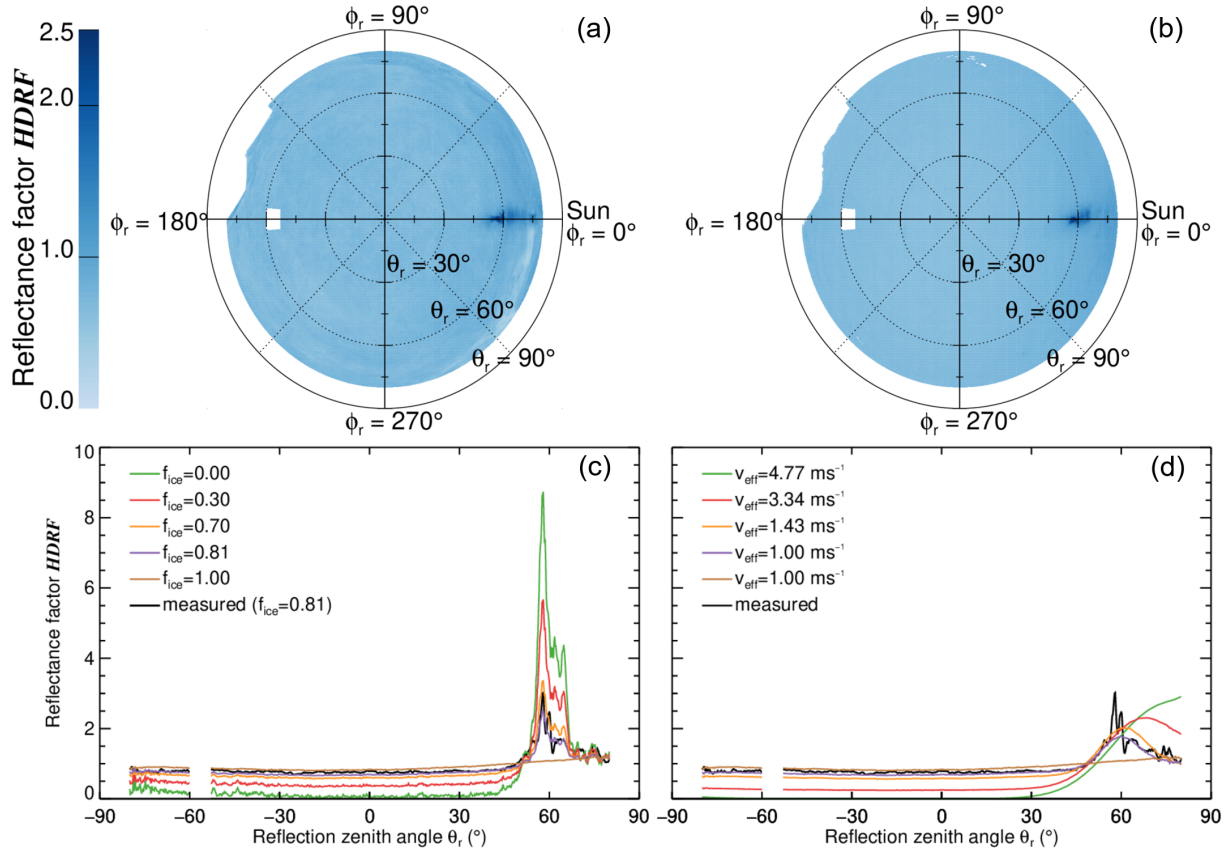
Figure 11. Polar plot of (a) the mean MIZ HDRF of the test data set (obtained by averaging over all images without separation), and (b) the constructed MIZ HDRF (calculated from the separated sea ice and open ocean HDRFs of the separation data set according to Eq. 5 using a sea ice fraction of 0.83). (c) Cross-section of the constructed MIZ HDRF along the solar principal plane obtained from the separated HDRFs of the separation data set for different sea ice fractions f_{ice} (color-coded) and the mean MIZ HDRF of the test dataset (black). (d) same as (c), but sea ice fraction-dependent simulations were used for the open ocean HDRF, see text for details.

test data set. It is obvious, that for each pixel the pixel-based sea ice fraction is different, covering a wide range between 0.5 and 1.0. ~~Sea ice fraction observed at each pixel throughout the entire image sequence (pixel-based sea ice fraction). Pixels which~~

455   ~~observed sea ice more often than on average (fraction > 0.83), have a larger mean MIZ HDRF compared to the reconstructed~~ ~~MIZ HDRF and vice versa. This pixel-wise distribution of the sea ice and open ocean surfaces of the individual images is still~~ ~~imprinted in the mean MIZ HDRF. A higher~~ Increasing the number of images ~~,~~ (which could be reached by, e. g., increasing the sampling frequency~~,~~ currently 1/6 Hz) ~~,~~ would reduce such effects and the pixel-based sea ice fraction would become more homogeneous.

460   Figures 11c and 11d show ~~the cross-section of the reconstructed~~ cross-sections of the constructed MIZ HDRF along the solar principal plane for different sea ice fractions~~, including the observed mean sea ice fraction of 0.83, and the mean MIZ~~
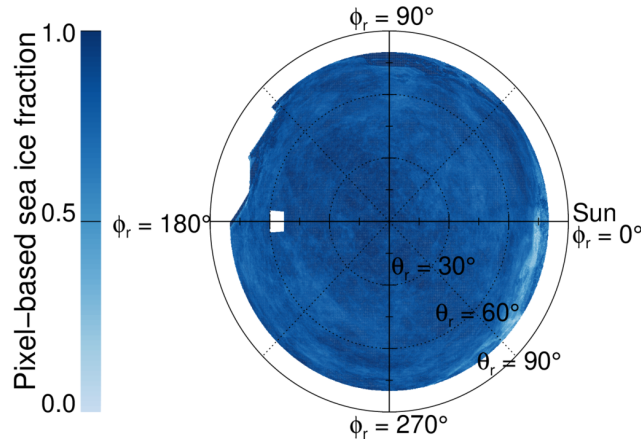
Figure 12. Sea ice fraction observed at each pixel throughout the images used for the test sata set (pixel-based sea ice fraction).

~~HDRF (black lines).~~ In Fig. 11c the individual sea ice and open ocean HDRFs separated from the separation data set are combined for different sea ice fractions. The open ocean and sea ice HDRFs are represented by the sea ice fractions of 0 and 1, respectively. While the ~~dependence of the open ocean HDRF on the~~ HDRF outside the sunglint increases with increasing

465   sea ice fraction ~~(Eq. 6) is neglected in~~, the sunglint contribution decreases without changing its shape. However, as shown in Fig. 7, there is a significant mismatch between the open ocean HDRF in the MIZ and that of the ice-free ocean due to wave attenuation. In Fig. 11c, the ~~sea ice fraction-dependent simulations are used~~ constructed HDRF for all sea ice fractions was retrieved using the MIZ open ocean HDRF (sea ice fraction of 0.83). To assess the impact of the surface roughness on the constructed HDRF, in Fig. 11d ~~Based on the linear combination (Eq. 5), both panels show a decrease of the magnitude of the~~

470   ~~sunglint peak with increasing~~, $R_{\mathrm{HDRF,ocean}}$ is replaced by a simulated HDRF that depends on the surface roughness varying with f_{ice}. The simulations are performed in the same way as described in Sect. 4.1 except that the input surface wind speed v_{eff} is parametrized as a linear function of the sea ice fraction. ~~However, in Fig. 11d, also the shift of the sunglint towards the horizon is visible as the effective wind speed increases with decreasing~~:

$$v_{\mathrm{eff}} = v_{\mathrm{meas}} \cdot (1 - f_{\mathrm{ice}}). \tag{6}$$

475   v_{eff} is considered as an effective wind speed, that would produce the same surface roughness and, thus, the same open ocean HDRF if the ocean was ice-free. v_{meas} is the wind speed measured at flight altitude and extrapolated to $10\,\mathrm{m}$ altitude. It has to be noted that this very basic relation between surface wind speed and sea ice fraction ~~. Furthermore, the maximum of the sunglint contribution is significantly decreased in Fig~~ aims only to illustrate the effects in a qualitative view. Numbers may change if observations for different sea ice conditions are considered. The comparison of Figs. 11 ~~d. For sea ice fractions lower~~

480   ~~than observed, this is partly due to the higher~~ c and 11d reveals significant differences in the shape and the position of the sunglint. With decreasing sea ice fraction, the increasing effective wind speed ~~. Additionally, the sharp peak that is visible in~~ (Eq. 6) causes a shift of the sunglint towards the horizon (compare Fig. 7 in Fig. 11d. Furthermore, the irregular shape and sharp peak visible in Fig. 11c ~~as a result of the pure specular reflection observed in some images,~~ is not present in the

~~simulations, which reduces the HDRF maximum. This difference in the sunglint region is the major discrepancy between the~~ ~~reconstructed MIZ HDRF for the observed~~ smooth simulations. This leads also to a reduced HDRF maximum in Fig. 11d. Nevertheless, the comparison of either of the constructed HDRFs (Figs. 11c or 11d) for the sea ice fraction ~~(0.83~~observed in the test dataset (0.81, purple line~~in~~) to its mean MIZ HDRF (black line) reveals only small differences outside the sunglint. The HDRF constructed with the simulated open ocean HDRF (Fig. 11d) ~~and~~ even shows good agreement to the mean MIZ HDRF ~~. Outside the sunglint , they show good agreement. For 65~~at the sunglint slope. In total, the difference between both HDRFs is less than 0.1 for about 82 % of the directions of the solar principle plane~~the relative difference between both HDRFs is within the uncertainty of the HDRF measurements. ~~.

This analysis ~~indicates~~ has shown that the linear construction of the HDRF ~~in the MIZ~~ from individual HDRFs of ~~sea ice~~ ~~and open ocean weighted by the sea ice fraction is possible. However, the impact of the sea ice distribution on the surface~~ ~~roughness and, thus,~~ open ocean and sea ice is well applicable if the environmental conditions are considered correctly. That also includes the parametrization of the surface roughness of the open ocean ~~HDRF in the MIZ has to be considered.~~ (effective surface wind speed), which considerably depends on the sea ice distribution. Neglecting those effects can lead to substantial irregularities in the resulting sunglint position and intensity

6   Conclusions

Reflected radiance measurements ~~from~~ were collected by an airborne 180~~°~~° fish-eye camera ~~were used to analyze the surface~~ ~~HDRF~~ in the MIZ north of Svalbard in June 2017. ~~The~~ From these data, the HDRF was calculated during cloud-free conditions for a 20-minute sequence of 138 camera images covering different sea ice fractions. The HDRFs of sea ice and open ocean surfaces were separated by applying a sea ice mask with different reflectivity and color ratio thresholds.

From the separated images, the averaged HDRFs of open ocean and partly snow-covered sea ice surfaces in the MIZ were derived. ~~The number of pixels used for the separated open ocean HDRF was rather low, which led to a higher variability~~ ~~compared to the separated sea ice HDRF. Nevertheless, the separated HDRFs~~ They confirmed the general features of ~~the~~ open ocean and sea ice ~~HDRFs~~ reported in literature (e. g., Warren et al., 1998; Gatebe et al., 2003; Jackson and Alpers, 2010). However, a comparison ~~of the separated open ocean HDRF~~ with simulations indicated that the common BRDF parametrizations for homogeneous open ocean surfaces as function of the wind speed (Cox and Munk, 1954) partly differ in the MIZ. This is ~~because, firstly, the~~ mainly due to wave attenuation between the ice floes in the MIZ (Kohout et al., 2011) ~~leads~~ leading to a reduced surface roughness compared to a homogeneous open ocean surface with the same surface wind speed. ~~Thus, a narrower~~ ~~and more intense sunglint is present in the MIZ, which may approach pure specular reflection. The irregular-shaped~~ This effect narrows the sunglint and intensifies its magnitude. The irregular shape of the sunglint observed in the ~~separated open ocean~~ ~~HDRF~~ data set was a result of the limitation of the sunglint mask and the highly variable surface roughness associated with the irregular distribution of sea ice and open ocean in the MIZ. ~~Secondly, the reflectance outside the sunglint could be enhanced in~~ ~~the MIZ due to horizontal photon transport from the sea ice surfaces.~~

The separated ~~sea ice HDRF was compared with independent measurements from the literature . It was found to be lower than that of homogeneous surfaces. This could be due to horizontal photon transport (Schäfer et al., 2015), complementary to the open ocean HDRF. However, since other unknown properties (such as snowgrain shape or surface roughness) affect the snow HDRF, the difference could also originate from the large variety of these properties.~~ HDRF of partly snow-covered sea ice ranged between independent literature HDRFs of homogeneous snow and bare ice surfaces. However, the comparison also revealed the large diversity of snow/sea ice HDRF patterns associated with the variability of snow and ice properties. Minor differences between the HDRF in the MIZ and that of homogeneous surfaces could originate as a result of the radiative effects of the contrasting surface type nearby (e. g., Ricchiazzi and Gautier, 1998; Schäfer et al., 2015).

The ~~mean~~ averaged HDRF of the MIZ ~~, calculated as average of all 138 images of the 20-minute section without separation,~~ showed features of ~~the HDRFs of both the~~ both sea ice and ~~the~~ open ocean surfaces. Even for rather high sea ice fractions, there is still a contribution from the sunglint in the MIZ~~which needs to be considered in~~, which might affect the analysis of ~~observations~~satellite observations in these reflection angles. This especially holds ~~regarding the permanent~~true with respect to ~~the~~ presence of leads in Arctic (e. g., Ivanova et al., 2016).

The mean MIZ HDRF of a subset of the analyzed data set was compared to the ~~reconstructed MIZ HDRF~~constructed one, calculated as a linear combination of the separated HDRFs of the remaining subset weighted by the sea ice fraction. The comparison showed good agreement for the measured sea ice fraction with a difference of less than ~~6 % for 86~~0.1 for 84 % of the pixels. ~~The reconstructed MIZ HDRF appeared more smoothed than the mean MIZ HDRF since the calculation of the reconstructed MIZ HDRF assumes the~~ Due to the assumption of a directionally constant sea ice fraction~~to be constant for all reflection directions. In total~~, the constructed HDRF of the MIZ was found to be smoother than the mean MIZ HDRF.

Altogether, this analysis implies that the ~~combination of homogeneous HDRFs is possible~~construction of the MIZ HDRF from individual sea ice and open ocean HDRFs provides meaningful results. This approach ~~becomes even more relevant as for most applications only the average sea ice fraction but not the exact distribution of~~ could become relevant for randomly distributed sea ice and open ocean~~in the MIZ~~, where only the sea ice fraction is known.

However, ~~it was shown that~~ the impact of the ~~ice floe distribution on the surface roughness and, thus, the sunglint~~ wave attenuation on the open ocean HDRF in the MIZ has ~~to be considered for the choice of the open ocean HDRF. To demonstrate these effects, a simple approach was proposed which parametrized the effective wind speed as a function of the sea ice fraction. The improved results indicated, that observations or simulations for homogeneous surfaces need to be treated with caution when applied in the MIZ.~~

~~This study showed that airborne observations of HDRF, even if limited to a short flight section, can help to characterize the surface reflection properties of inhomogeneous surfaces~~a significant impact also on the sunglint pattern of the mean MIZ HDRF. This effect needs to be considered in retrieval methods similar to the one used here. To improve the applicability of such methods, further research is needed, regarding the parametrization of the surface roughness of open ocean in the MIZ. ~~Similar measurements in~~Also the impact of the exact floe distribution on the surface reflectance properties needs to be investigated further. To extend the method to different environmental conditions (e. g., sea ice fraction, surface wind speed)~~are needed to~~

550 ~~fully parametrize~~ , further measurements are needed for a full parametrization of the HDRF in such complex scenarios ~~which likely will dominate the Arctic in a future climate.~~ as the MIZ, which may be the dominant surface type of the future Arctic.

Data availability. All data measured by the research aircraft *Polar 5* during ACLOUD and used in this study are published on the PANGAEA database. The radiances measured by the digital camera equipped with fish-eye lens can be found at Jäkel and Ehrlich (2019, https://doi.org/10.1594/PANGAEA.901024). The spectral irradiance data measured by SMART were published by Jäkel et al. (2019, https://doi.org/10.1594/PANGAEA.899177). The wind speed was published together with other meterological parameters (Hartmann et al., 2019, https://doi.org/10.1594/PANGAEA.902849). The radiosounding used for radiative transfer simulations is available in Maturilli (2017, https://doi.org/10.1594/PANGAEA.879822) . The HDRF data described by Goyens et al. (2018) can be found on SEANOE (Goyens et al., 2015, https://doi.org/10.17882/55352).

Author contributions. All authors contributed to the discussion of the results and the editing of the article. SB selected the case study, analyzed the data and drafted the article. SB developed the sea ice mask and performed the radiative transfer simulations. AE initiated the study. EJ processed the radiance and irradiance data. MW and AE designed the experimental basis of this study.

Competing interests. The authors declare that they have no conflict of interest.

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