



# Intercomparison of snowfall estimates derived from the CloudSat Cloud Profiling Radar and the ground-based weather radar network over Sweden

L. Norin<sup>1</sup>, A. Devasthale<sup>1</sup>, T. S. L'Ecuyer<sup>2</sup>, N. B. Wood<sup>3</sup>, and M. Smalley<sup>2</sup>

<sup>1</sup>Atmospheric Remote Sensing Unit, Research Department, Swedish Meteorological and Hydrological Institute, Norrköping, Sweden

<sup>2</sup>Department of Atmospheric and Oceanic Sciences, University of Wisconsin-Madison, Madison, Wisconsin, USA

<sup>3</sup>Cooperative Institute for Meteorological Satellite Studies, University of Wisconsin-Madison, Madison, Wisconsin, USA

Correspondence to: L. Norin (lars.norin@smhi.se)

Received: 10 July 2015 – Published in Atmos. Meas. Tech. Discuss.: 6 August 2015

Revised: 17 November 2015 – Accepted: 20 November 2015 – Published: 1 December 2015

**Abstract.** Accurate snowfall estimates are important for both weather and climate applications. Ground-based weather radars and space-based satellite sensors are often used as viable alternatives to rain gauges to estimate precipitation in this context. In particular, the Cloud Profiling Radar (CPR) on board CloudSat is proving to be a useful tool to map snowfall globally, in part due to its high sensitivity to light precipitation and its ability to provide near-global vertical structure. CloudSat snowfall estimates play a particularly important role in the high-latitude regions as other ground-based observations become sparse and passive satellite sensors suffer from inherent limitations.

In this paper, snowfall estimates from two observing systems – Swerad, the Swedish national weather radar network, and CloudSat – are compared. Swerad offers a well-calibrated data set of precipitation rates with high spatial and temporal resolution, at very high latitudes. The measurements are anchored to rain gauges and provide valuable insights into the usefulness of CloudSat CPR's snowfall estimates in the polar regions. In total,  $7.2 \times 10^5$  matchups of CloudSat and Swerad observations from 2008 through 2010 were intercompared, covering all but the summer months (June to September). The intercomparison shows encouraging agreement between the two observing systems despite their different sensitivities and user applications. The best agreement is observed when CloudSat passes close to a Swerad station (46–82 km), where the observational conditions for both systems are comparable. Larger disagreements outside this range suggest that both platforms have difficulty

with shallow snow but for different reasons. The correlation between Swerad and CloudSat degrades with increasing distance from the nearest Swerad station, as Swerad's sensitivity decreases as a function of distance. Swerad also tends to overshoot low-level precipitating systems further away from the station, leading to an underestimation of snowfall rate and occasionally to missing precipitation altogether. Several statistical metrics – including the probability of detection, false alarm rate, hit rate, and Pierce's skill score – are calculated. The sensitivity of these metrics to the snowfall rate, as well as to the distance from the nearest radar station, are summarised. This highlights the strengths and the limitations of both observing systems at the lower and upper ends of the snowfall distributions as well as the range of uncertainties that can be expected from these systems in high-latitude regions.

## 1 Introduction

Snowfall is a crucial component of the Earth's water and energy cycle (Levizzani et al., 2011; Waliser et al., 2011). Its effect on weather and climate are multi-faceted over high-latitude regions. At shorter timescales, ranging from days to months, snowfall can readily change the surface temperature, impact atmospheric dynamics, and influence circulation patterns. Thus, accurate representation of snowfall is one of the key challenges confronted by forecasting and numerical

weather prediction models. Characterising snowfall at sub-daily to daily scales is of great societal value. For example, heavy snowfall caused by a convective snow-band event can completely blanket the transportation infrastructure, leading to traffic chaos. For high-latitude countries like Sweden, the timely information on snowfall helps in planning daily communal services as well as to better manage tourism and agricultural industries. Ground-based weather radars are most commonly used to monitor precipitation for such weather applications.

From the climate perspective, better understanding of snowfall is also warranted (Waliser et al., 2011). For example, variability in snowfall directly influences variations in surface albedo and temperature and thus has a profound impact on surface radiation balance. Snow cover also impacts surface–air interactions by regulating heat and mass exchanges. Snowfall further has far-reaching impacts via teleconnections. For example, the variability in Eurasian snow cover has been shown to influence Asian monsoon rainfall (Liu and Yanai, 2002), and changes in snow cover onset over Siberia have been reported to influence Southeast Asian monsoon rainfall (Ye et al., 2005). Other studies argue that the increased snowfall over the Himalayas leads to a reduction in Indian monsoon rainfall (Turner and Slingo, 2011), mainly due to increased reflection and cooling of the surface, leading to weakened land–sea thermal contrast, which is considered an important trigger for the strength of monsoonal circulation. During spring, melting snow regulates the surface and river run-off in catchments. Hence long-term changes in snowfall characteristics directly impact the hydrological cycle at a regional scale.

The need for snowfall measurements over large regions is often met by using ground-based radar systems. Space-based remote sensing is often used, either as a viable alternative or in combination with ground-based measurements, to monitor precipitation for climate applications. Microwave imagers on board polar-orbiting satellites, for example, have contributed greatly to our understanding of the global precipitation systems in the last few decades. Data from space-borne instruments are an important component of the Global Precipitation Climatology Project (GPCP; Huffman et al., 2009), while NASA's Global Precipitation Measurement (GPM) suite offers combined observations from a microwave imager and a dual-frequency precipitation radar, which have potential for snowfall studies (Hou et al., 2014). Geostationary satellites also help in monitoring precipitation, but for a very high latitude country like Sweden their usefulness is limited due to the high viewing and relative azimuth angles, especially in the northern parts of Sweden.

In spite of its importance, characterising snowfall globally has been difficult due either to the absence or limited coverage or due to the limited capabilities of the observing systems. In situ measurements from precipitation gauges provide invaluable data, but they are mainly restricted to land areas and are geographically inhomogeneous and sparse, es-

pecially over high-latitude regions. Apart from networks of ground-based weather radars, covering very small geographical areas in both hemispheres, a reliable source of snowfall information over the high latitudes and the polar regions (where snowfall matters most) is generally lacking. Space-based observation of snowfall is, in a broader sense, in its infancy. The satellite sensors that operate at microwave or millimetre wavelengths are showing promise to obtain quantitatively reliable estimates (Noh et al., 2009; Surussavadee and Staelin, 2009; Levizzani et al., 2011; Liu and Seo, 2013), but they are still of inadequate quality over the high latitudes, especially over land and ice-covered surfaces. Optical imaging sensors do provide information on the snow cover extent, but the critical information on snowfall rate still remains illusive. This is, however, changing since the launch of CloudSat, as a part of the A-Train convoy of satellites, in 2006 (L'Ecuyer and Jiang, 2010). For the first time, the active Cloud Profiling Radar (CPR) on board CloudSat offers a possibility of obtaining realistic estimates of snowfall rate from space (Liu, 2008; Kulie and Bennartz, 2009; Wood, 2011; Wood et al., 2013).

As every observing system has its strengths and weaknesses, the estimates of snowfall rates from both the CloudSat CPR and ground-based weather radar systems could be improved if the synergy of these platforms could be exploited (Cao et al., 2014; Smalley et al., 2014). With their better spatial and temporal resolution and calibration against high-quality rain gauges, ground-based weather radars offer an independent source of information to intercompare snowfall estimates from the CloudSat CPR. On the other hand, the snowfall estimates from CloudSat CPR, unlike those from weather radars, are uniformly calibrated. Furthermore, ground-based radar beams can overshoot shallow precipitating systems, missing them as the observation distance from the radar increases. But even though CloudSat is generally insensitive to the vertical location of the precipitating system, the lowermost few bins of CloudSat, ranging roughly from 600 to 1200 m in height, can also be affected by ground clutter (Maahn et al., 2014). With regard to detectability, CloudSat is about an order of magnitude more sensitive to very light precipitation than any other existing space-based sensor (Skofronick-Jackson et al., 2012). On the other hand, unlike weather radars, CloudSat can saturate in the case of heavy-snowfall events (Cao et al., 2014). All of these considerations suggest that quantifying the strengths and weaknesses of these two observing systems (i.e. CloudSat CPR and ground-based weather radars) would not only be beneficial to improve their snowfall estimates but would also help in bracketing the spread in their expected uncertainties. Such knowledge could certainly help in evaluating snowfall variability simulated by climate models.

In a previous work Cao et al. (2014) compared CloudSat's snowfall detectability and estimation of snowfall rate over the contiguous United States to that of the snowfall estimate from the National Multi-Sensor Mosaic quantitative

precipitation estimate (NMQ) (Zhang et al., 2011). The study found that CloudSat has good detectability for light snowfall but underestimates moderate and heavy snowfall, due to attenuation. In another study Smalley et al. (2014) compared all precipitation measurements (rain as well as snow) from CloudSat to that of the NMQ. They found that the NMQ observes significantly fewer precipitation events than CloudSat and that the performance of the NMQ, compared to CloudSat, decreases for near-surface temperatures below 0 °C. Evidently, both these observing systems have their strengths and weaknesses. In order to further evaluate CloudSat's performance it is therefore important to compare its measurements to other ground-based data sets.

The main focus of the present study is to quantitatively intercompare snowfall estimates from CloudSat and the ground-based weather radar network over Sweden. The Swedish radar network has been operational since the 1980s, and more than a decade of archived precipitation data exist to intercompare with space-based estimates (Michelson et al., 2000; Michelson and Koistinen, 2002; Michelson, 2006; Devasthale and Norin, 2014). Apart from offering an independent source of snowfall estimates, the high-latitude geographical position of the Swedish radars entails sampling of different meteorological regimes under which CloudSat CPR data could be intercompared. The next section provides an overview of CloudSat CPR and the Swedish weather radar data set, followed by a description of the results in the third section. The results are summarised in the final section.

## 2 Data sets and processing

### 2.1 Snowfall product 2C-SNOW-PROFILE from CloudSat

CloudSat snowfall estimates were obtained from the 2C-SNOW-PROFILE data product, release R04 (Wood, 2011; Wood et al., 2013). This product uses a Bayesian optimal estimation retrieval algorithm (Rodgers, 2000) to estimate vertically resolved properties of snowfall from vertical profiles of W-band (94 GHz) reflectivities measured by CloudSat's CPR. The CloudSat orbit is such that the radar makes observations between 82° N and 82° S latitude, completing one orbit approximately every 99 min (Tanelli et al., 2008) and repeating its orbital ground track every 16 days. This orbit leads to moderately dense spatial sampling at high latitudes. Profiles have a horizontal spatial resolution of 1.7 km along-track by 1.4 km cross-track. The radar has an intrinsic vertical resolution of 485 m, but measurements are oversampled to provide an effective vertical resolution of 239 m.

Rather than assuming a fixed relationship between reflectivity and snowfall rate (a so-called  $Z$ - $S$  relationship), the 2C-SNOW-PROFILE retrieval algorithm estimates vertical profiles of the probability density functions (PDFs) of snow particle size distribution parameters. These posterior PDFs are estimated by minimising a cost function that incorporates

a priori estimates of the environmental distributions of these parameters as well as uncertainty-weighted differences between the observed and forward-modelled radar reflectivity profile. The reflectivity forward model uses high-quality descriptions of the PDFs of snow particle microphysical and scattering properties as functions of size (Wood et al., 2015) as well as treatments for attenuation and multiple scattering by snow particles. The estimates of the size distribution parameters are then used along with the forward model's microphysical properties to construct the vertically resolved PDFs of snowfall rate and other snow properties.

A retrieval is performed if the 2C-PRECIP-COLUMN product (Haynes et al., 2009) has categorised the surface precipitation as snow or as mixed-phase with a melted mass fraction of less than 10 %. The melted mass fraction is estimated based on the height of the freezing level and assumptions about the environmental lapse rate below the freezing level. Temperature information is obtained from reanalysis products of the European Centre for Medium-Range Weather Forecasting co-located to the CloudSat profile.

The surface snowfall rate is estimated from the retrieved profile of snowfall rates. Ground clutter affects the CPR measurements in the radar range bins nearest the surface, so these near-surface bins cannot be included in the reflectivity profiles when retrievals are performed, creating what is sometimes called a blind zone. Over land, this zone extends about 1 km above the surface. Consequently, the approach taken currently by 2C-SNOW-PROFILE is to estimate the surface snowfall rate as the rate retrieved in the radar bin immediately above the blind zone. This surface snowfall rate is assigned a confidence value from "none" to "high" depending on the expected performance of the forward model and the reliability of the temperature-based estimate of the precipitation phase, among other factors.

### 2.2 Snowfall product based on the Swedish weather radar network

The Swedish weather radar network consists of 12 horizontally polarised C-band Doppler radars. These radars measure three spectral moments: reflectivity, radial velocity, and spectrum width. From these moments quantities such as precipitation rate, wind speed, and turbulence are estimated. In this work we focus only on reflectivity as our main interest is the snowfall rate.

Reflectivity,  $Z$ , measures the fraction of returned power and is interpreted in terms of the backscattering characteristics of the observed particles. The Swedish radars output reflectivities within  $-30 \leq Z \leq 71.6$  dBZ in steps of 0.4 dBZ. The minimum reflectivity value,  $Z_{\min}$ , is assigned to all measurements ranging from  $-\infty$  to  $-30$  dBZ. Such measurements are referred to as undetected measurements and are interpreted by the radar as conditions with no precipitation. As the strength of an echo decreases with the square of its distance, further from the radars the minimum detectable sig-

nal increases (see, e.g., Doviak and Zrnić, 2006) which also leads to an increase in the upper limit of  $Z_{\min}$ .

To suppress ground echoes, the Swedish weather radars are equipped with clutter filters. The clutter filters work by omitting the amplitudes of the three frequency channels closest to zero in the frequency spectrum, suppressing echoes with radial velocities less than  $\pm 1 \text{ m s}^{-1}$ . The radar receiver is protected from overload by damping nearby signals by 60 dB, making data from the first 4 km from the radar unusable.

The scan strategy of the Swedish radars consists of performing azimuthal scans around a vertical axis for 10 different tilt angles,  $\theta$ . The lowest scan is made at  $\theta = 0.5^\circ$ , and the highest is made at  $\theta = 40^\circ$ . These scans, which together make up a polar volume data set, are repeated every 15 min. Relevant radar characteristics are summarised in Table 1.

From polar volume data sets horizontal cross sections of radar reflectivity at a certain altitude can be generated. Over areas where no data exist at the specified altitude the measurement nearest in height is selected. Such cross sections are referred to as pseudo-constant altitude plan position indicator (PCAPPI). In Sweden, the PCAPPIs are defined at 500 m altitude above the corresponding radar.

Nordrad (Carlsson, 1995) is a close collaboration among the Swedish Meteorological and Hydrological Institute (SMHI); the Norwegian Meteorological Institute; the Finnish Meteorological Institute; the Estonian Environment Agency; and the Latvian Environment, Geology and Meteorology Agency. An additional agreement exists with the Danish Meteorological Institute. Within the Nordrad collaboration radar PCAPPIs are exchanged in real time. All together, there are currently 35 weather radars operating in Sweden, Norway, Finland, Estonia, Latvia, and Denmark.

At the SMHI, composite radar images covering the Nordic countries are generated using PCAPPI data from as many available weather radars as possible. Before merging the PCAPPIs into a composite image, some quality adjustments are made. Radar measurements may be affected by nearby obstacles such as trees or mountains. A beam blockage correction, based on the method by Bech et al. (2003), is applied to correct for the reduction in reflectivity due to topography. Radar measurements may also contain echoes from non-precipitating objects such as ground clutter or clear-air targets. Non-precipitation echoes are removed by a filter using satellite observations. Radar echoes from areas which are classified as cloud free by the satellite are removed by the filter (Michelson, 2006).

Due to a combination of the positive non-zero angle of the lowest scan and the curvature of the Earth's surface, measurements at increasing distances from the radars correspond to increasing heights (with increasing risk of completely or partially overshooting precipitating clouds); see, e.g., Fig. 2. Furthermore, since the beam widens with distance, the measurement volume becomes larger, which increases the risk for incomplete beam filling. As a result, precipitation es-

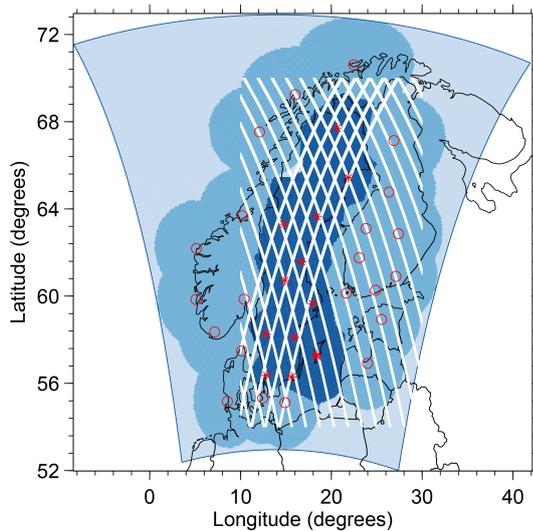
**Table 1.** Selected characteristics of the Swedish weather radars.

Tilt angles	0.5, 1.0, 1.5, 2.0°	2.5, 4.0, 8.0, 14.0, 24.0, 40.0°
Transmit power	250 kW	250 kW
Wavelength	5.35 cm	5.35 cm
Gain	44.7 dB	44.7 dB
Pulse width	0.5 $\mu\text{s}$	0.5 $\mu\text{s}$
Beam width	0.9°	0.9°
PRFs	600/450 Hz	1200/900 Hz
Rotational speed	2 rpm	2 rpm
Measurement radius	240 km	120 km
Radial resolution	2 km	1 km
Azimuthal resolution	0.86°	0.86°
Range cells	120	120
Azimuth gates	420	420
Max unambiguous velocity	24 $\text{m s}^{-1}$	48 $\text{m s}^{-1}$

timates at large distances from the radar may be of lower quality than those within 100 km from the radar. In order to correct for this distance dependence, data from rain gauges, which are considered to measure precipitation accurately, are used to calibrate the PCAPPIs. In order to compare radar reflectivities to the rain-gauge-measured precipitation, the reflectivities are converted to precipitation rate  $R$  ( $\text{mm h}^{-1}$ ) using the empirical relationship  $Z = aR^b$ , where  $a = 200$  and  $b = 1.5$  (see, e.g., Battan, 1973).

The radar data are adjusted to the rain gauges by fitting a second-degree polynomial to the logarithmic gauge-to-radar ratio, as a function of distance to the nearest radar. Radar measurements above 0.1 mm and rain gauge measurements above 0.5 mm from 1 week's worth of time, to ensure that sufficiently many data pairs exist, are used in the calculations. The gauge adjustment method is described in detail in Michelson and Koistinen (2002). The quality-adjusted PCAPPIs are used to produce a composite image, Nordrad, covering the Nordic countries. The Nordrad composite image has a spatial resolution of  $2 \text{ km} \times 2 \text{ km}$  and is generated every 15 min.

It should be noted that Nordrad assumes a fixed  $Z$ – $R$  relationship, regardless of the phase of the precipitation, even though a large variety of  $Z$ – $S$  relationships have been suggested (see, e.g., Rasmussen et al., 2003). Compared to one such  $Z$ – $S$  relationship,  $Z = 70S^2$ , the relationship used by Nordrad ( $Z = 200R^{1.5}$ ) underestimates snowfall intensity for  $Z < 35 \text{ dBZ}$  and overestimates snowfall for  $Z > 35 \text{ dBZ}$ . For example, a radar echo of 10 dBZ corresponds to a precipitation rate of  $0.14 \text{ mm h}^{-1}$  using Nordrad's  $Z$ – $R$  relationship, whereas the same echo strength, using  $Z = 70S^2$ , corresponds to a precipitation rate of  $0.38 \text{ mm h}^{-1}$ . However, the error introduced in Nordrad by using a fixed  $Z$ – $R$  relationship is to a large extent corrected by the adjustment to rain gauges. It should also be pointed out that Nordrad does not apply a vertical profile of reflectivity (see, e.g., Germann and Joss, 2002) to correct its data. Even though the

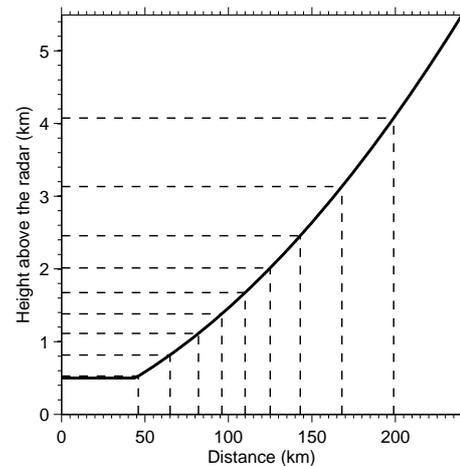


**Figure 1.** Area covered by the Nordrad composite (light blue). Coverage by all Nordic radars is shown in blue. Data originating from Swedish radars (Swerad) are displayed in dark blue. The positions of the Swedish radars are depicted by red stars, whereas the locations of the other radars are shown by red circles. Tracks of the selected CloudSat passes are illustrated by white lines.

rain gauge adjustment corrects for the range dependence, the large reflectivities occasionally caused by melting snow (so-called bright-band effects) are not treated, which can sometimes lead to an overestimation of precipitation. For a recent study of the performance of Nordrad, see Berg et al. (2015).

For the purpose of this study, a simple but robust snowfall product was generated by combining Nordrad's precipitation composite with 2 m temperatures from SMHI's system for operational mesoscale univariate analyses of selected meteorological parameters (MESAN; Häggmark et al., 2000). The model domain of MESAN covers Scandinavia and the entire drainage basin of the Baltic Sea. The analysis is performed on a rotated latitude–longitude grid with a spatial resolution of  $0.1^\circ$  (11 km) and a time resolution of 1 h. For every Nordrad composite image the 2 m temperature field from MESAN nearest in time and space was used to generate a snowfall product. If the 2 m temperature was less than or equal to  $0^\circ\text{C}$ , the corresponding precipitation from Nordrad was classified as snow, otherwise as rain.

In this work we have used snowfall estimates from the Nordrad composites but only selected measurements originating from the area covered by Swedish radars. For the purpose of this paper we call this selection of data Swerad. Figure 1 shows the area covered by Nordrad and Swerad together with the location of all radars as well as the selected tracks of CloudSat.



**Figure 2.** Measurement height as a function of distance for Swerad (thick, solid line). Thin, dashed lines show the range limits that were used to divide the Swerad data into 10 range rings with an equal number of observations (0–46, 46–65, 65–82, 82–96, 96–110, 110–125, 125–143, 143–168, 168–199, and  $> 199$  km).

### 2.3 Processing

During 2007 a modification was made to the Swedish radar hardware to implement Doppler processing for all scans. In order to have a homogeneous data set, we have therefore only used radar data collected after 1 January 2008, even though CloudSat was launched in 2006. For the present study, we used all ascending and descending passes of CloudSat (except during summer months, June to September) from 1 January 2008 to 31 December 2010 between  $54$  and  $70^\circ$  N latitude bands. In total, we analysed 1143 tracks, and the maximum number of matchups for any track was 1741.

For every CloudSat pass over Sweden, its coordinates and time were extracted and the Swerad radar pixels nearest in time and space were selected. If the distance to an observation by CloudSat was larger than 2 km or if the difference in time was more than 7.5 min, the data were discarded. Furthermore, only CloudSat observations with a confidence flag corresponding to “moderate” or “high” were analysed. For every CloudSat pass 77–1741, co-located observations were collected. In total, 716 545 observational pairs were collected and analysed.

As described in Sect. 2.2, measurements from the ground-based radars generally have lower quality far away from the radar station. Figure 2 shows the height above the radar as a function of distance. In order to examine the effect of the varying measurement height, we have divided Swerad's measurements into 10 range rings, each containing approximately the same number of measurements (i.e.  $70 \times 10^3$ ). These range rings (0–46, 46–65, 65–82, 82–96, 96–110, 110–125, 125–143, 143–168, 168–199, and  $> 199$  km) are also shown in Fig. 2.

It should be noted that frontal systems, which are mainly responsible for snowfall over Sweden, could advect by as much as 5 km in the maximum 7.5 min co-location time difference allowed between CloudSat and Swerad. The average temporal difference is, however, below 4 min, and even though this time difference can introduce some uncertainty while comparing the snowfall retrievals, especially if any one of the observing system leads or lags the passing front, the impact on the results of this study is expected to be small.

In the Supplement, Fig. S1 shows a histogram of the 2 m temperatures from MESAN, corresponding to the snowfall measurements reported by CloudSat. From Fig. S1 it is seen that more than 90 % of CloudSat's measurements correspond to 2 m temperatures below 0 °C and that the most frequent temperature was approximately −2 °C. CloudSat and Swerad therefore seem to agree well on when snowfall occurs.

### 3 Results and discussion

In order to get an overview of how the snowfall retrievals from CloudSat and Swerad compare, we first investigated their empirical cumulative distribution function (ECDF). The ECDF is defined as

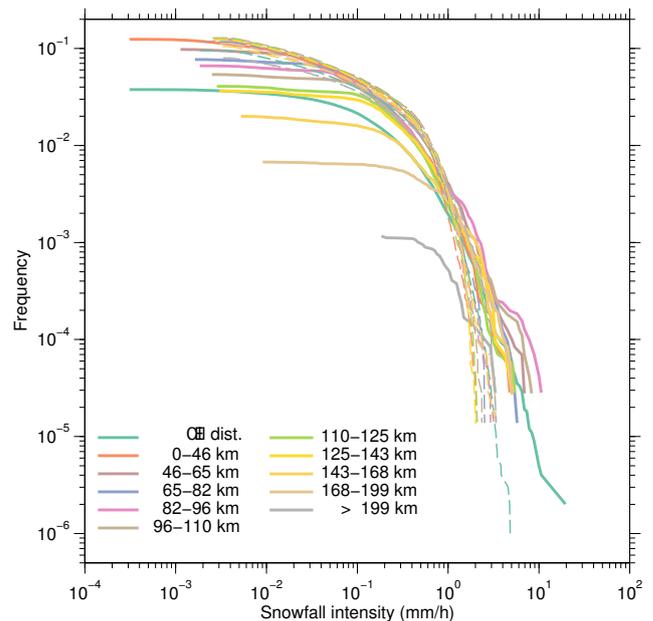
$$P(x) = 1 - \int_x^{\infty} p(x') dx', \quad (1)$$

where  $p(x)$  is the probability that a measurement has the value  $x$  and  $P(x)$  is the probability that a measurement has a value equal to or greater than  $x$ . In Fig. 3 ECDFs are shown for the observations made by CloudSat and Swerad. In addition to the ECDF for all distances, Fig. 3 shows the corresponding functions for the various range rings defined in Sect. 2.3.

From Fig. 3 it can be seen that the CloudSat ECDFs are, as expected, unaffected by the distance to the nearest ground-based radar. During all the co-located passes, CloudSat measured snow approximately 10 % of the time. A snowfall rate,  $S$ , higher than  $0.1 \text{ mm h}^{-1}$  was measured less than 5 % of the time. For heavier snowfall,  $S > 1 \text{ mm h}^{-1}$ , a rapid decrease in the frequency of occurrence can be seen. Only 2 % of CloudSat's observations estimated a snowfall rate  $S > 1 \text{ mm h}^{-1}$ .

The ECDFs from Swerad, on the other hand, show a larger variation depending on the distance to the nearest radar. The differences are most clearly seen for light-snowfall rates,  $S < 0.1 \text{ mm h}^{-1}$ . The greater the distance to the nearest radar, the larger the lowest estimated snowfall rate. The frequency of observed snowfall is also seen to be lower for increasing distances.

At large distances to the nearest radar ( $d > 199 \text{ km}$ ) snowfall was only detected in 1 % of the observations, whereas close to the radars ( $0 \leq d \leq 46 \text{ km}$ ) snow was reported over 10 % of the time. For intermediate snowfall rates,  $0.1 \leq$



**Figure 3.** Empirical cumulative distribution functions for observations of snowfall intensities from Swerad (solid lines) and CloudSat (dashed lines), for various distances to the nearest ground-based radar.

$S \leq 1 \text{ mm h}^{-1}$ , and small to medium distances to the nearest radar,  $d < 125 \text{ km}$ , the ECDFs of CloudSat and Swerad are similar. However, for large snowfall rates,  $S > 1 \text{ mm h}^{-1}$ , Swerad recorded more frequent and larger snowfall rates compared to CloudSat. The largest snowfall rate was estimated by Swerad as  $S \approx 20 \text{ mm h}^{-1}$ , whereas the largest snowfall rate detected by CloudSat was estimated near  $S = 5 \text{ mm h}^{-1}$ . It is interesting to note that Swerad reported snowfall rates larger than  $5 \text{ mm h}^{-1}$  for almost all distances to the nearest radar. As non-precipitation echoes (i.e. ground clutter or clear-air returns) are mainly expected to occur close to the radars, it seems that these large snowfall rates originate from precipitation and not from clutter. This suggests either that Swerad overestimates the snowfall rate for large reflectivities or that the current 2C-SNOW-PROFILE algorithm may have limited ability to retrieve heavy precipitation. Since Swerad assumes a fixed  $Z$ - $R$  relationship, as described in Sect. 2.2, it is expected that Swerad may overestimate snowfall rates for high reflectivity values. High reflectivity values can occur as a result of bright-band effects, which are not corrected in Swerad (see Sect. 2.2). On the other hand, as reported by Cao et al. (2014), CloudSat may also underestimate heavy snowfall due to attenuation or non-Rayleigh scattering effects that limit the sensitivity of W-band reflectivity to particle size in heavy snow.

In Fig. 3 we can further see that – except for short distances to the nearest radar,  $d < 65 \text{ km}$  – CloudSat observes snowfall more frequently than Swerad. This results partly from

Swerad overshooting shallow snowfall at large distances and from Swerad’s decrease in sensitivity for larger distances.

To compare the detection capabilities of each platform, the CloudSat and Swerad matchups were evaluated using the following metrics:

1. probability of detection (POD) for both snowfall and no snowfall (hereafter referred to as clear conditions),
2. false alarm rate (FAR) for both snowfall and clear conditions,
3. hit rate (HR),
4. Pierce’s skill score (PSS).

These quantities are defined as follows, using the notation in the contingency matrix in Table 2:

$$POD_{\text{snow}} = \frac{d}{c + d}, \quad (2)$$

$$POD_{\text{clear}} = \frac{a}{a + b}, \quad (3)$$

$$FAR_{\text{snow}} = \frac{b}{b + d}, \quad (4)$$

$$FAR_{\text{clear}} = \frac{c}{a + c}, \quad (5)$$

$$HR = \frac{a + d}{a + b + c + d}, \quad (6)$$

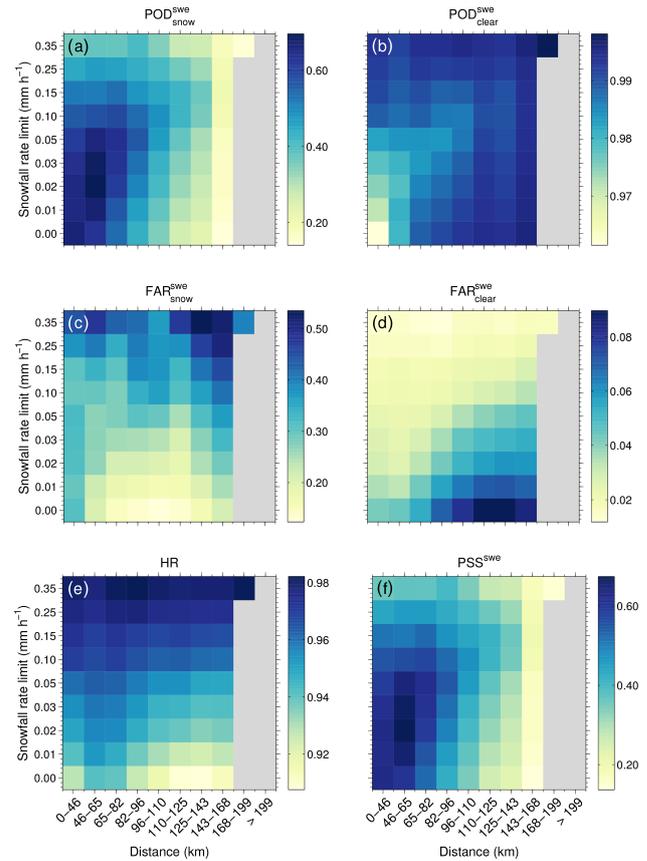
$$PSS = \frac{ad - bc}{(a + b)(c + d)}. \quad (7)$$

The POD and FAR estimate how efficient the evaluated system is in determining either snow or clear conditions. High values of POD as well as low values of FAR are expected for observational pairs that agree well. The HR measures the efficiency of the evaluated system’s ability to correctly classify clear or snowy conditions. As more than 90 % of the co-located observations originate in clear conditions, the HR score is expected to be high, even though the classification of snow might not be as good. The PSS score takes the uneven classification distribution into account and provides a more balanced measure of how well the evaluated system separated the snow from clear conditions.

The lower limit of detectable snowfall,  $S_{\text{lim}}$ , is ultimately determined by the sensitivity of the radar receiver. For CloudSat the lowest retrieved snowfall rate was  $S = 2 \times 10^{-3} \text{ mm h}^{-1}$ , whereas for Swerad the lowest detected snowfall varied between  $S = 3 \times 10^{-4}$  and  $S = 2 \times 10^{-1} \text{ mm h}^{-1}$ , depending on the distance to the nearest radar (cf. Fig. 3). In order to take the different sensitivities into account, we have calculated the statistical scores using different values for lower limit of detectable snowfall,  $S_{\text{lim}}$ : 0, 0.01, 0.02, 0.03, 0.05, 0.1, 0.15, 0.25, and  $0.35 \text{ mm h}^{-1}$ . As there are no error-free measurements to compare the snowfall estimates to (no “ground truth”), each score is calculated for both

**Table 2.** Contingency matrix for observations from CloudSat and Swerad.

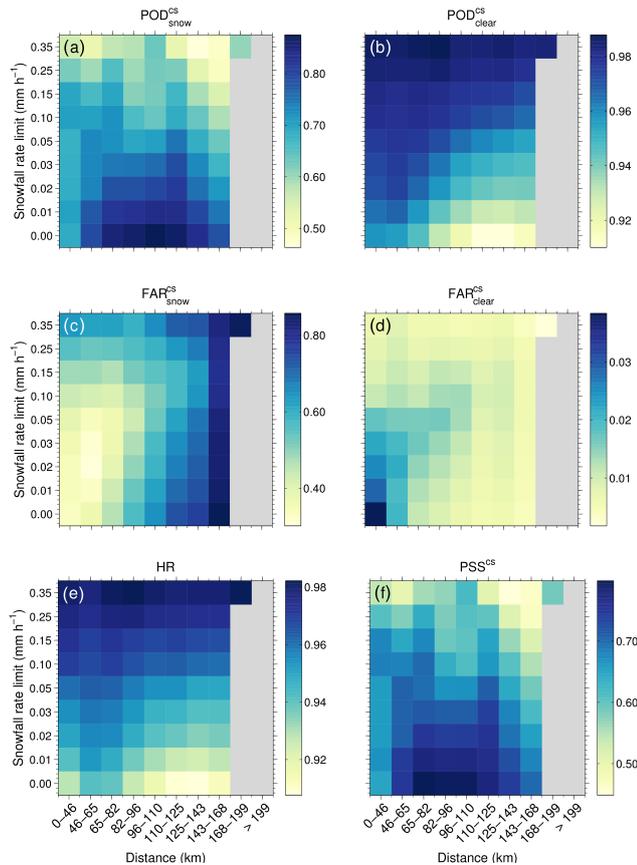
	Evaluated		
	Scenario	Clear	Snow
Reference	Clear	<i>a</i>	<i>b</i>
	Snow	<i>c</i>	<i>d</i>



**Figure 4.** Statistical scores for Swerad using CloudSat as a reference. The scores are shown as a function of distance to the nearest radar as well for different lower limits of the snowfall rate. The panels show the probability of detecting snow (a), the probability of detecting clear conditions (no precipitation) (b), false alarm rate of snow (c), false alarm rate of clear conditions (d), hit rate (e), and Pierce’s skill score (f).

CloudSat and Swerad using the other observing system as a reference.

Results of the skill scores are shown in Figs. 4 and 5. In Fig. 4 the observations by Swerad are evaluated using CloudSat as a reference, whereas in Fig. 5 CloudSat is evaluated using Swerad as a reference. Every skill score is shown as a function of distance to the nearest ground-based radar as well as a function of  $S_{\text{lim}}$ . Skill scores were only calculated when at least 150 observations in each category in the contingency matrix existed (cf. Table 2). A superscript on the



**Figure 5.** Statistical scores for CloudSat using Swerad as a reference. The scores are shown as a function of distance to the nearest radar as well for different lower limits of the snowfall rate. The panels show the probability of detecting snow (a), the probability of detecting clear conditions (no precipitation) (b), false alarm rate of snow (c), false alarm rate of clear conditions (d), hit rate (e), and Pierce's skill score (f).

skill scores (swe for Swerad and cs for CloudSat) denotes the evaluated system.

In Fig. 4a it is seen that the  $POD_{snow}^{swe}$  ranges from 0.15 to 0.7. The highest values of the  $POD_{snow}^{swe}$  are found for small distances to the nearest ground-based radar,  $46 < d < 65$  km, together with  $0.01 \leq S_{lim} \leq 0.05 \text{ mm h}^{-1}$ . For larger distances the  $POD_{snow}^{swe}$  decreases, and for  $d > 143$  km it approaches 0.1 for all values of  $S_{lim}$ . The decreasing values of the  $POD_{snow}^{swe}$  for increasing distances mean either that Swerad underestimates the frequency of snowfall or that CloudSat overestimates it. Since the altitude at which Swerad's measurements are made increases with increasing distance (cf. Fig. 2) while CloudSat exhibits a uniform beam height independent of range to the ground radar, it is likely that Swerad misses snowfall at large distances due to partial or complete overshooting. Furthermore, the sensitivity of the Swedish radars decreases with increasing distance, which

would also lead to a decrease in the  $POD_{snow}^{swe}$ , especially for low values of  $S_{lim}$ .

In Fig. 4a it is also seen that for increasing distances the  $POD_{snow}^{swe}$  obtains a maximum for increasing values of  $S_{lim}$ . This can be understood by examining the ECDFs presented in Fig. 3. The maximum value of the  $POD_{snow}^{swe}$ , for a given range ring, is found for the value of  $S_{lim}$  where the ECDFs of CloudSat and Swerad meet. For example, for  $96 < d < 110$  km the ECDFs meet for a snowfall intensity of  $0.1 \text{ mm h}^{-1}$ , and for this value of  $S_{lim}$  the  $POD_{snow}^{swe}$  obtains its maximum. For higher values of  $S_{lim}$  the  $POD_{snow}^{swe}$  is expected to decrease due to the variability in snowfall intensity in the observations from the evaluated system (Swerad) together with increasing the threshold for snowfall in the reference system (CloudSat).

The  $POD_{clear}^{swe}$ , shown in Fig. 4b, ranges from 0.96 to almost 1. Lower values of the  $POD_{clear}^{swe}$  are only found close to the radar,  $d < 46$  km, in light snow ( $S_{lim} \leq 0.02 \text{ mm h}^{-1}$ ). Close to the ground Swerad can detect non-precipitation echoes such as ground clutter or returns from clear-air targets. It is therefore expected that Swerad occasionally reports false precipitation at these distances. On the other hand, CloudSat can miss shallow snowfall that forms in its blind zone, as mentioned in Sect. 2.1. The  $POD_{clear}^{swe}$  increases for increasing  $S_{lim}$ , indicating that this effect is most pronounced for weak echoes or light snowfall.

Figure 4c shows the  $FAR_{snow}^{swe}$ , which ranges from 0.1 to 0.55. Except for very small distances to the nearest radar ( $d < 46$  km) the  $FAR_{snow}^{swe}$  is seen to increase for increasing values of  $S_{lim}$ . This is a result of selecting higher snowfall intensities from the evaluated system (Swerad) together with the variability in intensity of the snowfall observations from the reference system (CloudSat). However, nearest to the ground-based radar,  $d < 46$  km, the  $FAR_{snow}^{swe}$  obtains its minimum for  $S_{lim} = 0.1 \text{ mm h}^{-1}$ . This is again the result of Swerad reporting echoes from non-precipitation targets in combination with CloudSat missing snowfall in its blind zone.

The  $FAR_{clear}^{swe}$ , shown in Fig. 4d, is low,  $FAR_{clear}^{swe} < 0.04$ , close to the nearest ground-based radar, indicating that Swerad is good at detecting snowfall at this range. Further from the nearest radar the  $FAR_{clear}^{swe}$  increases up to 0.09 for  $S_{lim} = 0 \text{ mm h}^{-1}$ . However, for  $S_{lim} \geq 0.1 \text{ mm h}^{-1}$  the  $FAR_{clear}^{swe} < 0.03$  for all distances. This confirms the previously discussed suspicion that Swerad misses light snowfall at large distances due to partial overshooting or decreased sensitivity.

The HR is shown in Fig. 4e. The HR is seen to increase for increasing  $S_{lim}$ , from  $HR = 0.91$  to  $HR > 0.98$ . This occurs because the higher the  $S_{lim}$  the more observations are classified as clear by both CloudSat and Swerad. However, a more interesting relation is found as a function of distance to the nearest ground-based radar. Initially the HR increases with distance, reaching a maximum value for  $46 < d < 82$  km, be-

yond which it decreases monotonically. This result represents a combination of the previously discussed problems occurring close (non-precipitation echoes detected by Swerad and blind zone for CloudSat) and far (overshooting and decreased sensitivity by Swerad) from the nearest ground-based radar. The distance where HR obtains its maximum shows the optimum distance when comparing the intensity estimates from both systems. Referring to Fig. 2, it is encouraging to note that the radar beam is at an altitude of about 1 km at this range from the radar, coinciding with the height to which the CloudSat observations correspond.

Figure 4f shows Pierce's skill score,  $PSS^{swe}$ . This score varies within  $0.15 < PSS^{swe} < 0.7$ . The lowest values of the  $PSS^{swe}$  are found for large distances from the nearest ground-based radar, reflecting the low values of the  $POD_{snow}^{swe}$  seen in Fig. 4a. The highest values of the  $PSS^{swe}$  are found for small to moderate distances to the nearest radar ( $46 < d < 65$  km) together with low to moderate lower limits of the snowfall rate ( $0.01 \leq S_{lim} \leq 0.05 \text{ mm h}^{-1}$ ). This shows the distances and the values of  $S_{lim}$  for which the observations from the measurement systems agree best, when using CloudSat as a reference.

In Fig. 5 the same set of skill scores is shown but now using Swerad as the reference. However, not all skill scores are independent of those shown in Fig. 4. From the definitions of the skill scores (see Eqs. 2–7) and the contingency matrix (cf. Table 2) it can be seen that  $POD_{snow}^{swe/cs} = 1 - FAR_{snow}^{cs/swe}$  and  $POD_{clear}^{cs/swe} = 1 - FAR_{clear}^{swe/cs}$ .

The  $POD_{snow}^{cs}$  is shown in Fig. 5a. For low to moderate lower limits of the snowfall rate,  $0 \leq S_{lim} = 0.1 \text{ mm h}^{-1}$ , the  $POD_{snow}^{cs}$  ranges from 0.6 to 0.9, whereas for high lower limits of the snowfall rate,  $S_{lim} \geq 0.4 \text{ mm h}^{-1}$ , the  $POD_{snow}^{cs}$  decreases down to almost 0.45. The decreasing values of the  $POD_{snow}^{cs}$  for increasing values of  $S_{lim}$  are explained by the variability in the snowfall intensity of the evaluated system (CloudSat) together with the increasing threshold for snowfall of the reference system (Swerad). Close to the radars,  $d < 46$  km, CloudSat underestimates the frequency of snowfall reported by Swerad more than for larger distances, for  $S_{lim} \leq 0.1 \text{ mm h}^{-1}$ . This was also seen in Fig. 4b and is attributed to a combination of CloudSat missing shallow snowfall in its blind zone and Swerad reporting non-precipitation echoes.

The  $POD_{clear}^{cs}$ , shown in Fig. 5b, is very high, close to 1, except for  $d > 82$  km in combination with  $S_{lim} \leq 0.01 \text{ mm h}^{-1}$ , where  $POD_{clear}^{cs} < 0.94$ . The lower values of the  $POD_{clear}^{cs}$  are, as previously discussed, explained by Swerad missing or partially overshooting snowfall at large distances together with Swerad's decreased sensitivity for larger distances.

The  $FAR_{snow}^{cs}$  (see Fig. 5c) ranges from 0.3 to 0.9. The  $FAR_{snow}^{cs}$  is low close to the radars,  $d < 82$  km, together with  $S_{lim} \leq 0.1 \text{ mm h}^{-1}$ , indicating good agreement between the observing systems. For larger distances to the nearest ground-based radar the values of  $FAR_{snow}^{cs}$  decrease as a re-

sult of Swerad missing snowfall due to overshooting and decreased sensitivity.

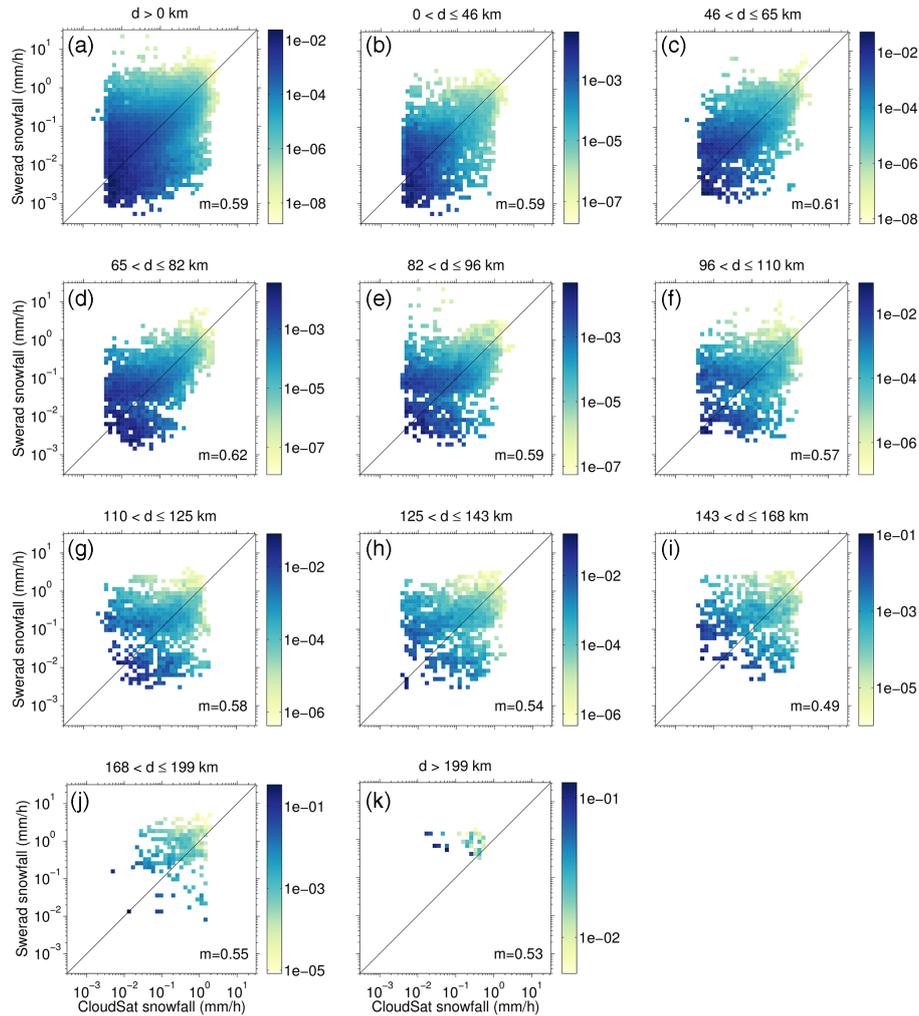
It can further be seen that for increasing distances the minimum value of the  $FAR_{snow}^{cs}$  is found for increasing values of  $S_{lim}$ . This can again be explained by a combination of Swerad missing observations of snowfall due to overshooting and decreased sensitivity and the variability in snowfall intensity in the observations from the reference system (Swerad) together with an increasing of the threshold of snowfall for the evaluated system (CloudSat).

Figure 5d shows the  $FAR_{clear}^{cs}$  which is almost zero everywhere, meaning that when CloudSat reports clear conditions Swerad almost never reports snow. Higher values,  $FAR_{clear}^{cs} > 0.02$  are only found close to the nearest ground-based radar,  $d < 65$  km, and for  $S_{lim} \leq 0.03 \text{ mm h}^{-1}$ . The higher values of the  $FAR_{clear}^{cs}$  confirm the previous observation that for these distances CloudSat measures snow less often than Swerad, due to CloudSat's blind zone and Swerad reporting non-precipitation echoes. Even very close to the nearest radar,  $d < 46$  km, the  $FAR_{clear}^{cs}$  approaches zero for increasing values of  $S_{lim}$ , suggesting that CloudSat predominantly underestimates light snowfall near the radars.

The hit rate, shown in Fig. 5e, is reproduced for convenience but is exactly the same as the hit rate shown in Fig. 4e since HR is symmetric with respect to the choice of reference. The PSS, however, differs depending on which system is used as a reference. The  $PSS^{cs}$ , using Swerad as a reference, is shown in Fig. 5f. This score varies within  $0.4 < PSS^{cs} < 0.8$ . The lowest values of the  $PSS^{cs}$  are found for high values of the lower limit of the snowfall rate limit,  $S_{lim} \geq 0.25 \text{ mm h}^{-1}$ , which reflects the corresponding low values of the  $POD_{snow}^{cs}$  (cf. Fig. 5a). For the highest value of the lower limit of the snowfall rate,  $S_{lim} = 0.35 \text{ mm h}^{-1}$ , the  $PSS^{cs}$  decreases for all distances and attains values between 0.4 and 0.6. High values of the  $PSS^{cs}$  are found for intermediate distances to the nearest radar ( $46 < d < 143$  km) together with  $S_{lim} \leq 0.03 \text{ mm h}^{-1}$ . This is a result of the corresponding high values of the  $POD_{snow}^{cs}$ . This shows again that an optimum range to the nearest ground-based radar exists where the problems close to (non-precipitation echoes from Swerad and CloudSat's blind zone) and far from (Swerad's decreased sensitivity and beam overshooting) the radars are minimised.

It is worth noting that these skill scores provide a clear demonstration of why, in the absence of a clearly superior data set, it is important to alternate the observing system used as the reference. The impact of different sensitivities, observational capabilities, and statistical artefacts resulting from these differences is clearly visible in the two sets of skill scores presented in Figs. 4 and 5.

The co-located observational pairs from CloudSat and Swerad are further analysed using 2-D PDFs in Fig. 6, which shows all observations when both CloudSat and Swerad reported snow. Ideally all co-located pairs should lie on the line of equality, but in reality, due to different sensitivities and



**Figure 6.** 2-D probability density functions (PDFs) of observational pairs of snowfall by CloudSat and Swerad. Panel (a) shows the 2-D PDF for all observational pairs, whereas panels (b–k) show the 2-D PDF for various distances to the nearest ground-based radar. Note the logarithmic scales. The metric  $m$  estimates how well the observations agree ( $0 \leq m \leq 1$ , where  $m = 1$  is perfect agreement).

observing principles, some scatter is expected. In order to provide an objective measure of the agreement between the observational pairs  $x_i$  and  $y_i$ , a simple metric, based on the normalised minimum distance of all  $N$  observational pairs to the line of equality, was defined:

$$m = 1 - \frac{1}{N} \sum_{i=1}^N \frac{\sqrt{(x_i - y_i)^2}}{x_i + y_i}, \quad (8)$$

where  $0 \leq m \leq 1$  and  $m = 1$  corresponds to all observational pairs lying on the line of equality.

In Fig. 6a the 2-D PDF for all observational pairs is shown. It can be seen that the most frequently observed snowfall rates were estimated as  $S < 0.1 \text{ mm h}^{-1}$  for both CloudSat and Swerad. The general agreement between the measurement systems is good,  $m = 0.59$ , and the agreement is even better for snowfall rates  $S > 0.1 \text{ mm h}^{-1}$ . For light-snowfall

rates,  $S < 0.05 \text{ mm h}^{-1}$ , Swerad tends to estimate slightly lower snowfall rates than CloudSat.

As shown in Fig. 3, the snowfall retrievals from Swerad are sensitive to the distance from the radar station. In order to investigate sensitivity to this distance, Fig. 6b–k show the 2-D PDFs for all observational pairs for different distances to the nearest ground-based radar (cf. Sect. 2.3 and Fig. 3). In Fig. 6b we see that close to the radar,  $d < 46 \text{ km}$ , the agreement between CloudSat and Swerad is good,  $m = 0.59$ . Very low snowfall rates,  $S < 0.05 \text{ mm h}^{-1}$ , tend to be estimated somewhat lower by Swerad. The same tendency is also seen for the next distance bin,  $46 < d < 65 \text{ km}$ , shown in Fig. 6c, but for this range to the nearest radar the overall agreement is higher,  $m = 0.61$ .

In Fig. 6d–k it is seen that, for increasing distance to the nearest radar, Swerad observes fewer and fewer light-snowfall events (cf. Fig. 3). This is likely a result of Swerad's

decreased sensitivity with increasing distance and by partial beam overshooting of the snowfall, discussed in Sects. 2.2 and 3. The agreement metric is seen to decrease from  $m = 0.62$  to  $m = 0.49$  (with the exception of slightly higher scores,  $m = 0.55$  and  $m = 0.53$ , for Fig. 6j and k). However, for snowfall rates  $0.1 \leq S \leq 1 \text{ mm h}^{-1}$  the agreement remains good, regardless of distance.

#### 4 Conclusions

In the present study, we exploited data from the Swedish ground-based radar network (Swerad) and the Cloud Profiling Radar on board the CloudSat satellite (2C-SNOW-PROFILE) to provide insights into their performance in snowing scenes. The different sensitivities and observing principles of these two systems offer increased understanding of their strengths and limitations. Furthermore, the high-latitude geographical location of Sweden allowed us to sample and intercompare snowfall retrievals under different meteorological and surface conditions than previously reported. In total, more than  $7.2 \times 10^5$  co-located observations were evaluated from 1 January 2008 till 31 December 2010, excluding the summer months.

The intercomparison shows encouraging agreement between the two observing systems. The distributions of snowfall rates are similar for CloudSat and Swerad for the range between  $0.1$  and  $1.0 \text{ mm h}^{-1}$ . The results from the sensitivity studies indicate that the best agreement is observed when CloudSat passes close to a Swerad station (46–82 km), where the observational conditions for both systems are comparable. Larger disagreements outside this range suggest that both platforms have difficulty with shallow snow, but for different reasons. We observed a clear tendency for the correlation between Swerad and CloudSat to degrade with increasing distance from the nearest Swerad station. This mainly occurs due to Swerad's decreased sensitivity for increasing distances but also as Swerad systematically overshoots low-level precipitating systems further away from the station, leading to an underestimation of the snowfall rate and occasionally to miss precipitation altogether. Data pairs close to the radar, on the other hand, suggest that some shallow-snow events that are detected by the ground-based radars are missed by CloudSat, due to ground clutter.

Several common performance metrics also show a clear tendency that the scores degrade with increasing distance from the Swerad stations. A closer inspection of the observations shows that the 2C-SNOW-PROFILE has limited ability to retrieve at the higher end of snowfall intensity distribution ( $> 1 \text{ mm h}^{-1}$ ), causing deviation from the corresponding distribution from Swerad. This is in agreement with the results found by Cao et al. (2014). On the lower end of the distribution, both observing systems seem to suffer from limitations. While Swerad detects non-precipitating, low-intensity echoes closest to the stations that are often misclassified as light snowfall, CloudSat CPR is also affected in the lower-

most kilometre from the surface, where its sensitivity is reduced considerably. These limitations make comparison of light-snowfall events difficult and impractical.

Finally, it should be mentioned that, although the design purpose and end users of these two observing systems are different, intercomparisons help to assess the performance of ground-based systems for weather applications, while also providing uncertainty information for climate applications that use satellite products (Boening et al., 2012; Palermé et al., 2014). The fact that Swerad and CloudSat CPR broadly agree with one another in the  $0.1$ – $1.0 \text{ mm h}^{-1}$  intensity range, recorded by the majority of snowfall events, could be exploited in future studies to reconcile differences between these two systems and in particular to improve the statistical relationship between reflectivity and snowfall rate derived from ground-based radars.

The results of this study suggest that clear improvements can be made to the Swerad data set. Correcting the ground-based radar data with a vertical profile of reflectivity would help eliminate bright-band effects and reduce some overestimated snowfall measurements. Furthermore, as Swerad does not use a  $Z$ – $S$  relationship for snowfall, there exists additional room for improvement. Formulating a representative  $Z$ – $S$  relationship has been one of the chronic problems, often discussed widely in the scientific community. CloudSat with its uniform calibration and beam height (together with information on cloud microphysics) has potential to be a realistic reference in this context.

**The Supplement related to this article is available online at doi:10.5194/amt-8-5009-2015-supplement.**

*Acknowledgements.* The portion of this research by T. S. L'Ecuyer, N. B. Wood, and M. Smalley was performed at the University of Wisconsin-Madison for the Jet Propulsion Laboratory, California Institute of Technology, sponsored by the National Aeronautics and Space Administration CloudSat programme.

Edited by: I. Moradi

#### References

- Battan, L. J.: Radar Observation of the Atmosphere, University of Chicago Press, Chicago, USA, 324 pp., 1973.
- Bech, J., Codina, B., Lorente, J., and Bebbington, D.: The sensitivity of single polarization weather radar beam blockage correction to variability in the vertical refractivity gradient, *J. Atmos. Ocean. Tech.*, 20, 845–855, doi:10.1175/1520-0426(2003)020<0845:TSOSPW>2.0.CO;2, 2003.
- Berg, P., Norin, L., and Olsson, J.: Creation of a high resolution precipitation data set by merging gridded gauge data and radar observations for Sweden, *J. Hydrol.*, doi:10.1016/j.jhydrol.2015.11, accepted, 2015.

- Boening, C., Lebsack, M., Landerer, F., and Stephens, G.: Snowfall-driven mass change on the East Antarctic ice sheet, *Geophys. Res. Lett.*, 39, L21501, doi:10.1029/2012GL053316, 2012.
- Cao, Q., Hong, Y., Chen, S., Gourley, J. J., Zhang, J., and Kirstetter, P. E.: Snowfall detectability of NASA's CloudSat: the first cross-investigation of its 2c-snow-profile product and national multi-sensor mosaic QPE (NMQ) snowfall data, *Prog. Electromagn. Res.*, 148, 55–61, doi:10.2528/PIER14030405, 2014.
- Carlsson, I.: NORDRAD – weather radar network, in: COST 75 Weather Radar Systems, edited by: Collier, C. G., European Commission, Luxembourg, 45–52, 1995.
- Devasthale, A. and Norin, L.: The large-scale spatio-temporal variability of precipitation over Sweden observed from the weather radar network, *Atmos. Meas. Tech.*, 7, 1605–1617, doi:10.5194/amt-7-1605-2014, 2014.
- Doviak, R. J. and Zrnić, D. S.: *Doppler Radar and Weather Observations*, 2 edn., Dover Publications, Mineola, New York, USA, 2006.
- Germann, U. and Joss, J.: Mesobeta Profiles to Extrapolate Radar Precipitation Measurements above the Alps to the Ground Level, *J. Appl. Meteorol.*, 41, 542–557, doi:10.1175/1520-0450(2002)041<0542:MPTEP>2.0.CO;2, 2002.
- Hägghmark, L., Ivarsson, K.-I., Gollvik, S., and Olofsson, P.-O.: Mesan, an operational mesoscale analysis system, *Tellus A*, 52, 2–20, doi:10.1034/j.1600-0870.2000.520102.x, 2000.
- Haynes, J. M., L'Ecuyer, T. S., Stephens, G. L., Miller, S. D., Mitrescu, C., Wood, N. B., and Tanelli, S.: Rainfall retrieval over the ocean with spaceborne W-band radar, *J. Geophys. Res.*, 114, D00A22, doi:10.1029/2008JD009973, 2009.
- Hou, A. Y., Kakar, R. K., Neeck, S., Azarbarzin, A. A., Kummerow, C. D., Kojima, M., Oki, R., Nakamura, K., and Iguchi, T.: The Global Precipitation Measurement Mission, *B. Am. Meteorol. Soc.*, 95, 701–722, doi:10.1175/BAMS-D-13-00164.1, 2014.
- Huffman, G. F., Adler, R. F., Bolvin, D. T., and Gu, G.: Improving the Global Precipitation Record: GPCP Version 2.1, *Geophys. Res. Lett.*, 36, L17808, doi:10.1029/2009GL040000, 2009.
- Kulie, M. and Bennartz, R.: Utilizing spaceborne radars to retrieve dry snowfall, *J. Appl. Meteorol.*, 48, 2564–2580, doi:10.1175/2009JAMC2193.1, 2009.
- L'Ecuyer, T. S. and Jiang, J.: Touring the atmosphere aboard the A-Train, *Phys. Today*, 63, 36–41, 2010.
- Levizzani, V., Laviola, S., and Cattani, E.: Detection and measurement of snowfall from space, *Remote Sensing*, 3, 145–166, doi:10.3390/rs3010145, 2011.
- Liu, G.: Deriving snow cloud characteristics from CloudSat observations, *J. Geophys. Res.*, 113, D00A09, doi:10.1029/2007JD009766, 2008.
- Liu, G. and Seo, E. K.: Detecting snowfall over land by satellite high-frequency microwave observations: the lack of scattering signature and a statistical approach, *J. Geophys. Res.-Atmos.*, 118, 1376–1387, doi:10.1002/jgrd.50172, 2013.
- Liu, X. and Yanai, M.: Influence of Eurasian spring snow cover on Asian summer rainfall, *Int. J. Climatol.*, 22, 1075–1089, doi:10.1002/joc.784, 2002.
- Maahn, M., Burgard, C., Crewell, S., Gorodetskaya, I. V., Kneifel, S., Lhermitte, S., Van Tricht, K., and Van Lipzig, N. P. M.: How does the spaceborne radar blind zone affect derived surface snowfall statistics in polar regions?, *J. Geophys. Res.-Atmos.*, 119, 13604–13620, doi:10.1002/2014JD022079, 2014.
- Michelson, D.: The Swedish weather radar production chain, in: *Proceedings of the Fourth European Conference on Radar in Meteorology and Hydrology*, Barcelona, Spain, 18–22 September 2006, 382–385, ERAD, 2006.
- Michelson, D. B. and Koistinen, J.: Gauge-radar network adjustment for the Baltic Sea Experiment, *Phys. Chem. Earth Pt. B*, 25, 10–12, 2002.
- Michelson, D. B., Andersson, T., Koistinen, J., Collier, C. G., Riedl, J., Szturc, J., Gjertsen, U., Nielsen, A., and Overgaard, S.: *BALTEX Radar Data Centre Products and their Methodologies*, Reports Meteorology and Climatology RMK 90, SMHI, Norrköping, Sweden, 76 pp., 2000.
- Noh, Y. J., Liu, G., Jones, A. S., and Vonder Haar, T. H.: Toward snowfall retrieval over land by combining satellite and in situ measurements, *J. Geophys. Res.-Atmos.*, 114, D24205, doi:10.1029/2009JD012307, 2009.
- Palmer, C., Kay, J. E., Genthon, C., L'Ecuyer, T., Wood, N. B., and Claud, C.: How much snow falls on the Antarctic ice sheet?, *The Cryosphere*, 8, 1577–1587, doi:10.5194/tc-8-1577-2014, 2014.
- Rasmussen, R., Dixon, M., Vasiloff, S., Hage, F., Knight, S., Vivekanandan, J., and Xu, M.: Snow Nowcasting Using a Real-Time Correlation of Radar Reflectivity with Snow Gauge Accumulation, *J. Appl. Meteorol.*, 42, 20–36, doi:10.1175/1520-0450(2003)042<0020:SNUART>2.0.CO;2, 2003.
- Rodgers, C.: *Inverse Methods for Atmospheric Sounding*, World Scientific Publishing, Singapore, 2000.
- Skofronick-Jackson, G. M., Johnson, B. T., and Munchak, S. J.: Detection thresholds of falling snow from satellite-borne active and passive sensors, *IEEE T. Geosci. Remote*, 51, 4177–4189, doi:10.1109/TGRS.2012.2227763, 2012.
- Smalley, M., L'Ecuyer, T., Lebsack, M., and Haynes, J.: A comparison of precipitation occurrence from the NCEP StageIV QPE Product and the CloudSat Cloud Profiling Radar, *J. Hydrol.*, 15, 444–458, doi:10.1175/JHM-D-13-048.1, 2014.
- Surussavadee, C. and Staelin, D. H.: Satellite retrievals of Arctic and equatorial rain and snowfall rates using millimeter wavelengths, *IEEE T. Geosci. Remote*, 47, 3697–3707, doi:10.1109/TGRS.2009.2029093, 2009.
- Tanelli, S., Durden, S. L., Im, E., Pak, K. S., Reinke, D. G., Partain, P., Haynes, J. M., and Marchand, R. T.: CloudSat's Cloud Profiling Radar after two years in orbit: performance, calibration, and processing, *IEEE Trans. Geosci. Remote*, 46, 3560–3573, 2008.
- Turner, A. G. and Slingo, J. M.: Using idealized snow forcing to test teleconnections with the Indian summer monsoon in the Hadley Centre GCM, *Clim. Dynam.*, 36, 1717–1735, doi:10.1007/s00382-010-0805-3, 2011.
- Waliser, D. E., Li, J.-L. F., L'Ecuyer, T. S., and Chen, W.-T.: The impact of precipitating ice and snow on the radiation balance in global climate models, *Geophys. Res. Lett.*, 38, L06802, doi:10.1029/2010GL046478, 2011.
- Wood, N. B.: Estimation of snow microphysical properties with application to millimeter-wavelength radar retrievals for snowfall rate, PhD thesis, Colorado State University, Fort Collins, CO, USA, available at: <http://hdl.handle.net/10217/48170> (last access: 3 August 2015), 2011.

- Wood, N. B., L'Ecuyer, T., Vane, D. G., Stephens, G. L., and Partain, P.: Level 2C snow profile process description and interface control document, Tech. rep., Colorado State University, available at: [http://www.cloudsat.cira.colostate.edu/sites/default/files/products/files/2C-SNOW-PROFILE\\_PDICD.P\\_R04.20130210.pdf](http://www.cloudsat.cira.colostate.edu/sites/default/files/products/files/2C-SNOW-PROFILE_PDICD.P_R04.20130210.pdf) (last access: 3 August 2015), 2013.
- Wood, N. B., L'Ecuyer, T. S., Heymsfield, A. J., and Stephens, G. L.: Microphysical constraints on millimeter-wavelength scattering properties of snow particles, *J. Appl. Meteorol. Clim.*, 54, 909–931, doi:10.1175/JAMC-D-14-0137.1, 2015.
- Ye, H., Bao, Z., and Feng, X.: Connections of Siberian snow onset dates to the following summer's monsoon conditions over Southeast Asia, *Int. J. Climatol.*, 25, 1567–1584, doi:10.1002/joc.1201, 2005.
- Zhang, J., Howard, K., Langston, C., Vasiloff, S., Kaney, B., Arthur, A., Van Cooten, S., Kelleher, K., Kitzmiller, D., and Ding, F.: National mosaic and multi-sensor QPE (NMQ) system: Description, results, and future plans, *B. Am. Meteorol. Soc.*, 92, 1321–1338, doi:10.1175/2011BAMS-D-11-00047.1, 2011.