

Regional and Seasonal Distribution of Arctic Low-Level Cloud Types and Their Relationship to Large-Scale En- 5 vironmental Conditions

Aymeric Dziduch¹, Guillaume Mioche^{1*}, Quentin Coopman², Clément Bazantay¹, Julien Delanoë³ and Olivier Jourdan^{1*}

10 ¹Université Clermont Auvergne, OPGC, Laboratoire de Météorologie Physique, 63000 Clermont-Ferrand, France

²Univ. Lille, CNRS, UMR 8518 - LOA - Laboratoire d'Optique Atmosphérique, F-59000 Lille, France

³LATMOS/IPSL, UVSQ Université Paris-Saclay, Sorbonne Université, CNRS, Guyancourt, France

Correspondence. Olivier Jourdan (Olivier.Jourdan@uca.fr)

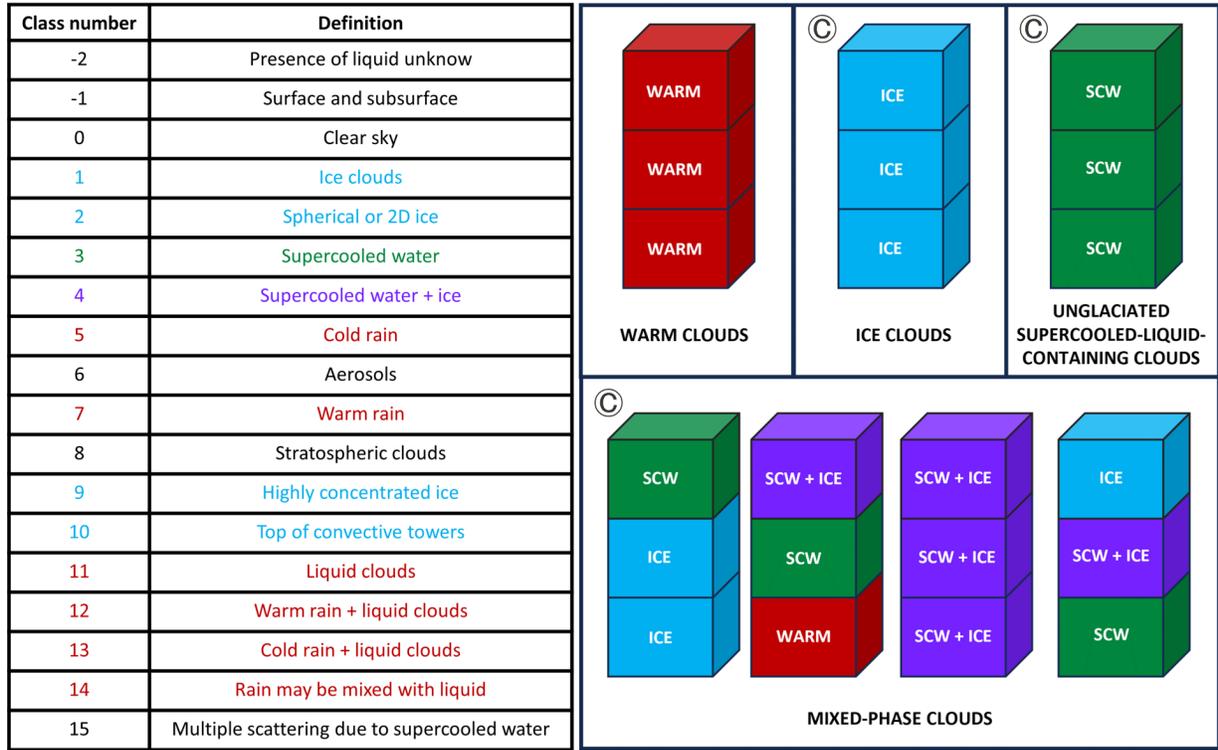
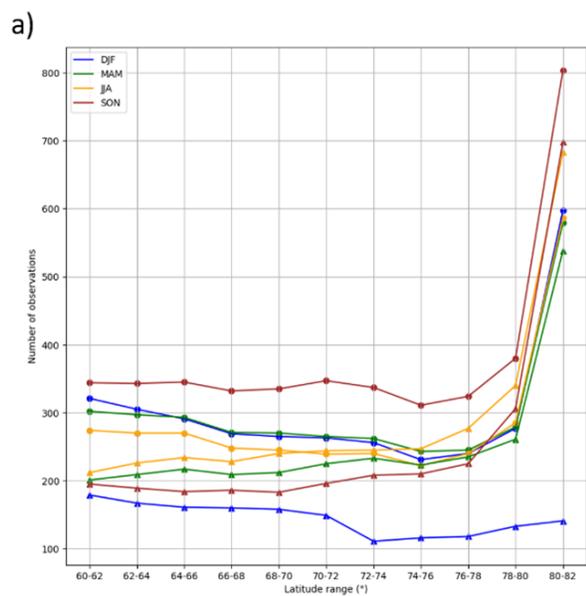


Figure S1. DARDAR classes used in the version V2.23 and cloud representation examples in DARDAR-SOCP (adapted from Bazantay et al., 2024). The colors of the classes in the table correspond to the pixel colors of the cloud representations. The symbol C in the diagrams represents cloud types belonging to the cold clouds (temperature < 0 °C).



b)

Year	Month											
	J	F	M	A	M	J	J	A	S	O	N	D
2007	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green
2008	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green
2009	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green
2010	Orange	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green
2013	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green
2014	Orange	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green
2015	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green	Green
2016	Orange	Orange	Orange	Green								

c)

	2007-2010		2013-2016	
	Total clouds	MPCs	Total clouds	MPCs
All seasons	51,39	18,87	51,44	17,72
DJF	42,29	12,89	43,71	12,63
MAM	47,11	15,78	49,06	16,41
JJA	46,58	18,58	48,04	17,72
SON	64,95	25,93	64,95	23,55

Figure S2. (a) Average number of observations per grid cell as a function of latitude for the 4 seasons (winter in blue, spring in green, summer in yellow, and autumn in brown). The circles represent the 2007-2010 period and the triangles the 2013-2016 period. Table (b) represents the availability of DARDAR mask v2.23 data for this study (green: available and orange: not available). Table (c) compares the average occurrences for the seasons between the 2 periods (2007-2010 and 2013-2016) for total clouds and MPCs.

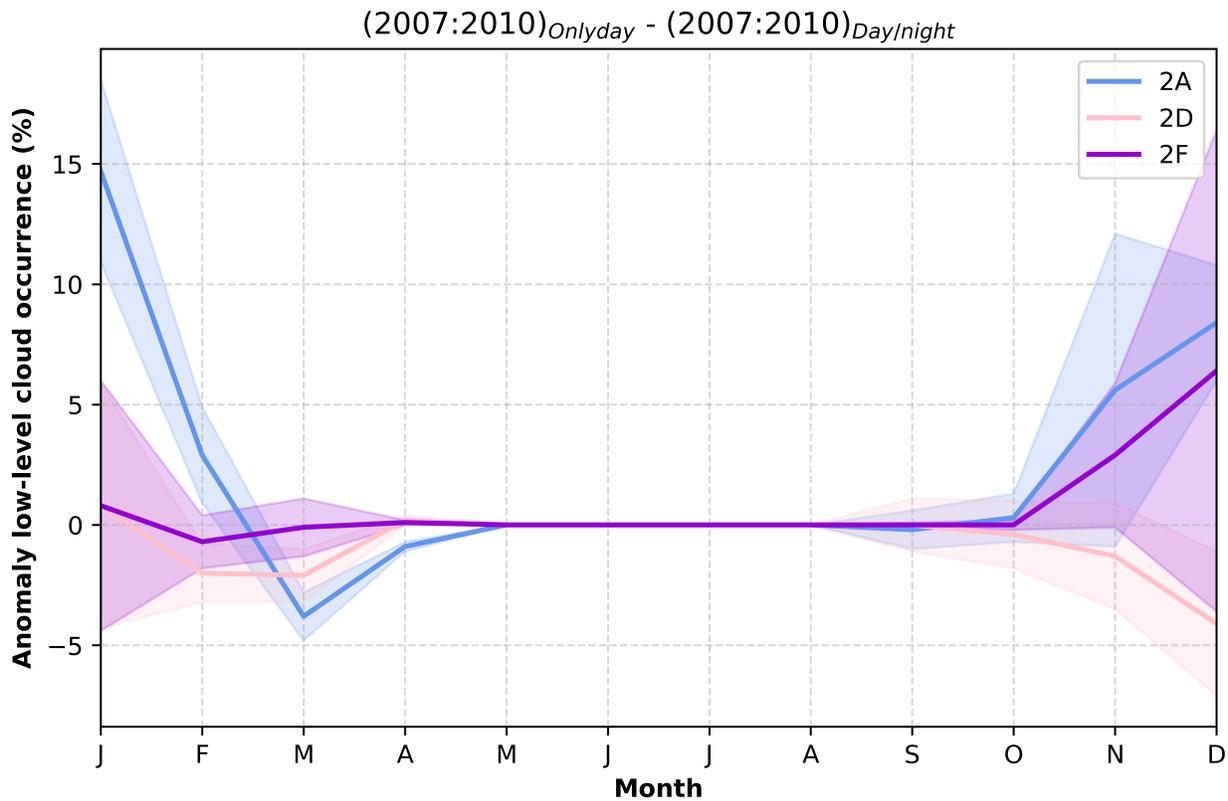


Figure S3. Cloud occurrence anomalies for 3 high-latitude regions (2A, 2D, and 2F) for the period 2007-2010 with two operating modes: day only and day/night. The colored curves represent the average values and the colored areas around the curves represent the standard deviation range.

a.

	SST	Surf_Temp	Surf_Pres	IWV	Temp_850hPa	Temp_700hPa	MCAO	LTS	EIS	SH_850hPa	SH_700hPa	GH_700hPa	GH_850hPa	Surf_Wind	Wind_850hPa	Wind_700hPa	SIC	AOD
MPCs	1,4	1,2	1,7	1,7	1,1	1,2	3,6	2,2	5,6	1,1	1,1	1,5	2,1	1,2	1,2	1,1	1,5	1,1
USLCs	1,1	1,1	1,6	1,3	1,3	1,3	1,9	1,3	1,1	1,5	1,6	1,5	1,2	1,1	1,3	1,2	1,0	1,1
Ice Clouds	1,8	2,1	1,1	1,3	12,4	8,5	1,4	1,3	1,1	1,5	1,6	1,2	1,5	1,3	1,2	1,2	1,0	1,1
Warm Clouds	3,6	2,8	1,3	1,2	8,4	5,6	1,5	1,1	1,7	3,3	2,7	1,1	2,0	1,0	1,0	1,1	1,6	1,0

Variance Inflation Factor (VIF)

b.

MPCs	0,27	0,18	0,42	0,42	-0,1	-0,2	-0,6	-0,6	-0,4	-0,1	-0,1	-0,3	-0,5	-0	-0,1	-0,1	-0,3	0,12
USLCs	-0,1	-0,1	0,37	0,25	-0,2	-0,2	-0,2	-0,2	-0,3	-0,3	-0,4	-0,2	-0,3	-0	-0,1	-0,1	-0	0,18
Ice Clouds	-0,5	-0,5	-0,1	0,08	-0,7	-0,7	-0,1	-0,2	0,18	-0,6	-0,6	-0,7	-0,4	-0	0,02	0,05	0,18	0,21
Warm Clouds	0,61	0,64	0,21	0,16	0,68	0,69	0,12	-0,1	-0,4	0,7	0,6	0,51	0,1	-0	-0	-0,1	-0,4	0,06

Spearman's rank correlations

Figure S4. (a) Variance Inflation Factor (VIF) computed for each explanatory variable and cloud type. Values exceeding the threshold of 10 indicate strong multicollinearity. (b) Spearman's rank correlations between cloud-type occurrences and environmental parameters.

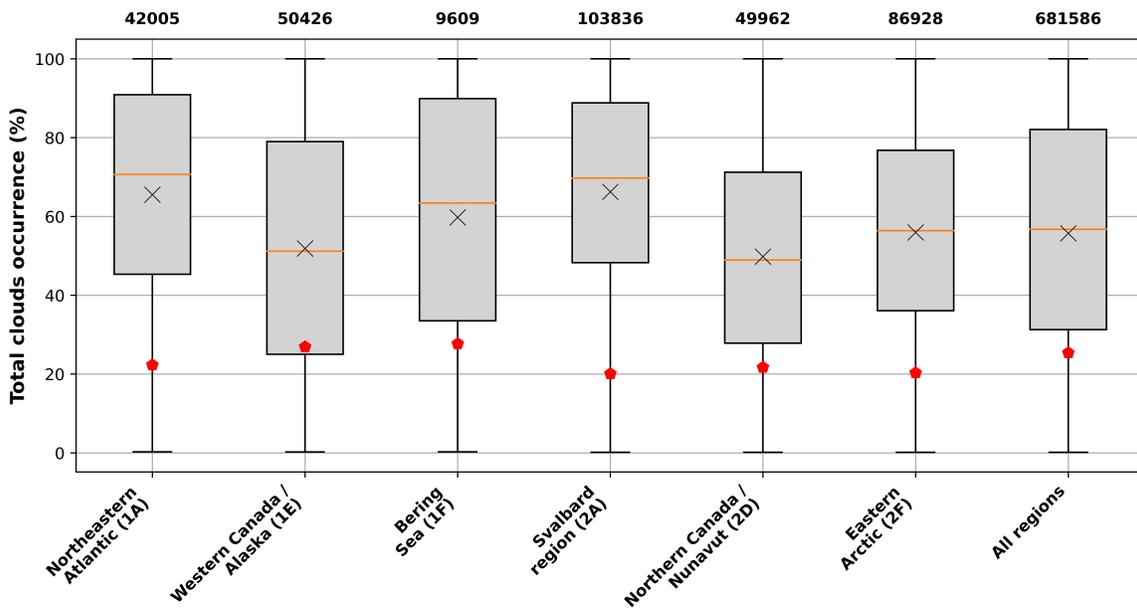


Figure S5. Boxplot showing the total cloud occurrence (%) statistics and uncertainties derived from DARDAR-MASK and DARDAR-SOCP products across different Arctic regions. The gray boxes represent the interquartile range (IQR), with the orange line indicating the median and the black cross representing the mean. Red pentagons denote the median absolute deviation (MAD). Numbers above each boxplot indicate the number of data points for each region. The final boxplot on the right merges data from all regions.

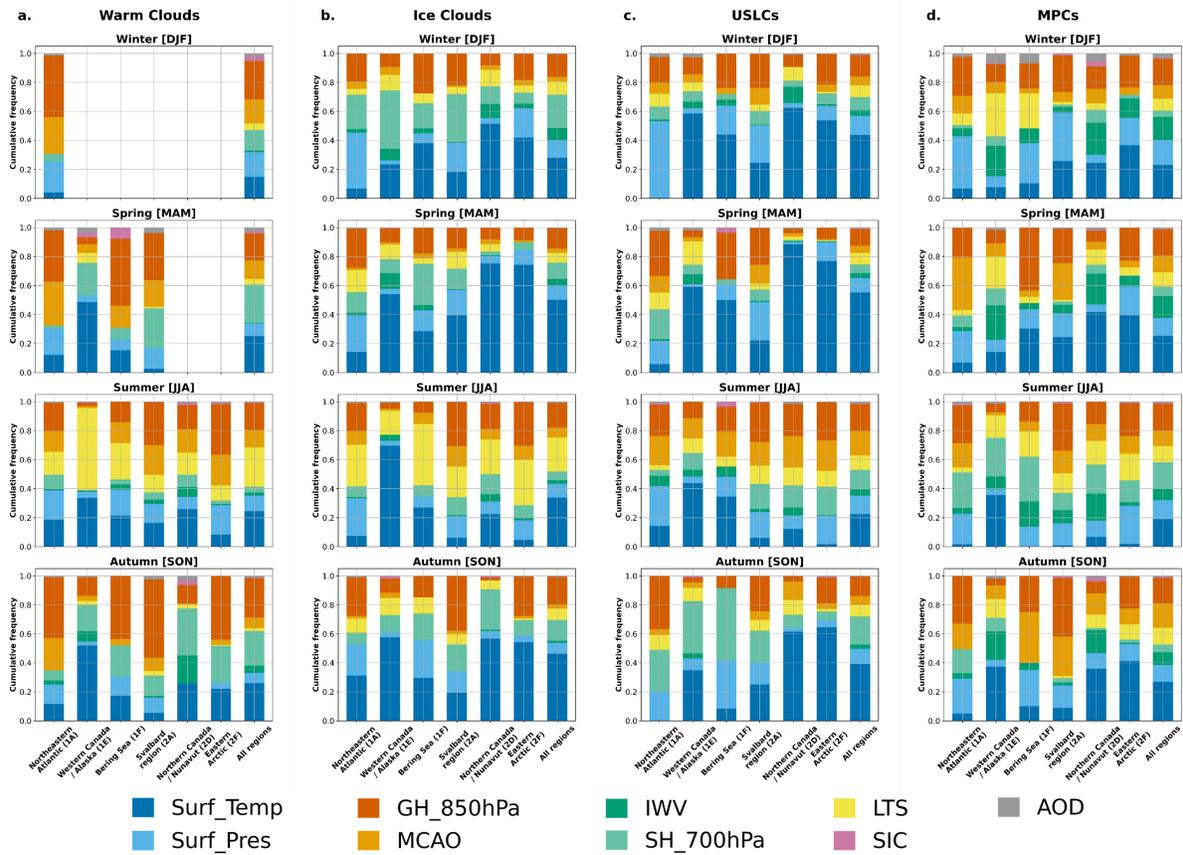


Figure S6. Cumulative frequency of most influential parameters of MLR for low-level (a) warm clouds, (b) ice clouds, (c) USLCs, and (d) MPCs. Each row corresponds to a season. The most influential parameter is only selected when MLR has $R^2 > 0.2$.

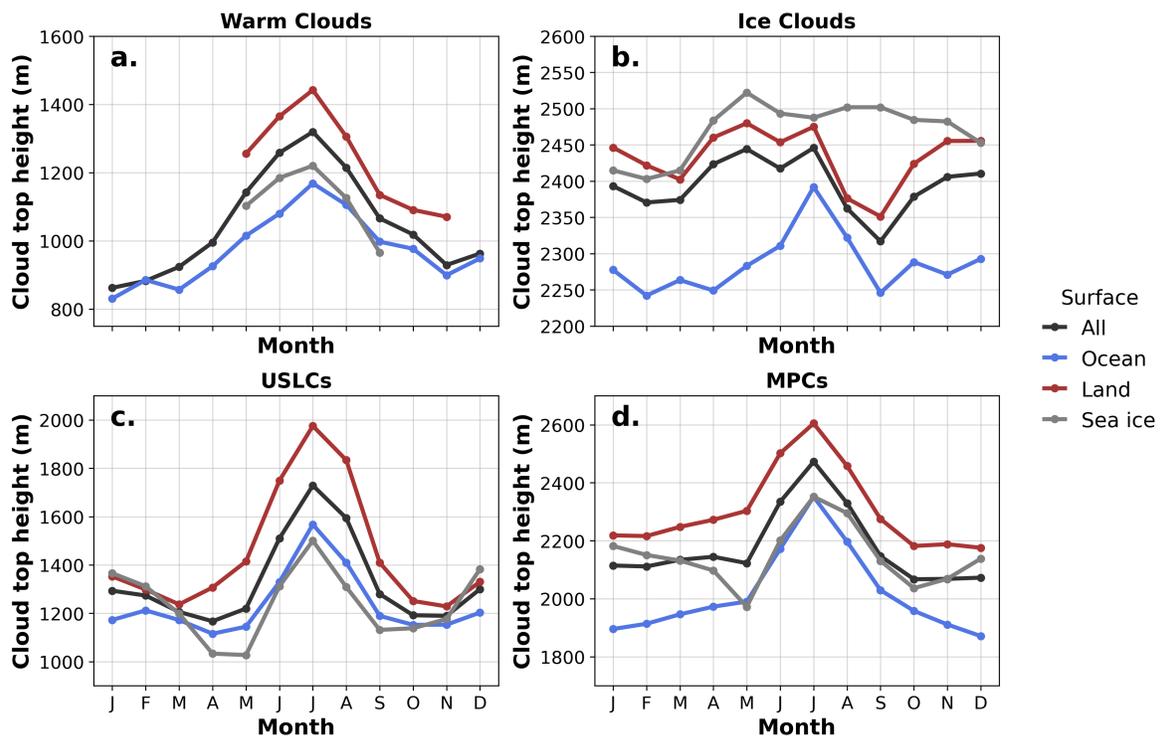


Figure S7. Monthly variations of the cloud-top height (m) for (a) warm clouds, (b) ice clouds, (c) USLCs, and (d) MPCs. Results are averaged over 2007–2016 and shown separately for different surface types: ocean (blue), land (red), sea ice (grey), and all surfaces combined (black).



Figure S8. Spearman's rank correlations, slopes, and p-value for three parameters: sea ice concentration (SIC), Lower-Tropospheric Stability (LTS), and surface temperature (Surf_Temp). Simple regressions are applied to the normalized data to compare the slopes. Correlations are made for MPCs, USLCs, ice clouds, and total clouds.

15 **Text S1.**

For the first 4 years of data (2007-2010), the CloudSat satellite operated normally, sampling the atmosphere 10 to 15 seconds ahead of CALIPSO. In 2011, a battery problem on CloudSat led to a loss of synchronization with CALIPSO and operational problems with the CPR radar. CloudSat resumed joint observations
20 with CALIPSO in 2012. However, technical problems necessitated the adoption of a new mode (Daylight Only Operations) for data from 2013 to 2016 (Braun et al., 2019; Listowski et al., 2019). During this second measurement period, CloudSat was only able to take atmospheric measurements during the daytime part of its orbit, for about 60 out of the 99 minutes of full orbit. As a result, about 40% of the scientific data could not be exploited (Kotarba, 2022). This missing data concerns the nighttime period. Cloud climatology
25 conducted using CloudSat data after 2012 shows a bias, particularly over the oceans (Noel et al., 2018). Data availability for this study is shown in Figure S2b.

A comparison of the average number of observations per grid cell as a function of latitude highlights the differences between seasons and periods (Fig. S2a). Overall, the number of observations is reduced by 30%
30 for low-latitude meshes (latitude $< 66-68^\circ$) between the two periods due to daylight-only operation in 2013-2016, especially for winter and autumn. For summer and spring, this difference is much less pronounced, due to the polar day during these seasons. For high-latitude grids (latitude $> 76-78^\circ$) the difference is very small in spring, summer, and autumn. Winter is the season with the largest discrepancy in the number of observations, especially for high-latitude meshes (latitude $> 76-78^\circ$) due to the polar night. The last mesh
35 (80-82°) loses up to 80% of the number of observations between the two periods. During the second period, observations are significantly reduced (especially in winter) for these meshes and, more generally, for high-latitude regions.

Figure S2c compares the average occurrences over the two periods (2007-2010 and 2013-2016). A comparison
40 between seasons shows that for total clouds, there seems to be no significant difference between the two periods. For example, there is a slight increase in average occurrences in winter for the period 2013-2016 (42.3% vs. 43.7%). This increase could be due both to the instrument malfunction and to the greater presence of clouds. The comparison for MPCs shows a slightly different behavior, with a seasonal decrease in occurrences over the period 2013-2016 (+1% for MAM vs. -2% for SON). Figure S3 shows a comparison of
45 occurrences for the period 2007-2010 (satellite instruments operational) using 2 operating modes (day only and day/night) for 3 high-latitude regions (affected by polar night). In terms of occurrences, the greatest differences are observed in December and January. However, these differences are larger for region Svalbard (2A) (Fig. S3) than for regions Eastern Arctic (2F) or Northern Canada/Nunavut (2D). The maximum difference for the Svalbard region (2A) is observed in January, with an average difference of 15% for all
50 low-level clouds. In contrast, Fig. S3 shows an underestimation of occurrences for the 2D region in early winter (November and December), but this remains below 5%. As in Noel et al. (2018), it seems that the strongest anomalies are found over oceanic areas. Comparisons for other cloud types show maximum differences of 5% average occurrence during January. These differences appear to be consistently positive for daytime measurements, which means that there may be slight overestimation of cloud occurrences due to
55 the CloudSat Daylight Only Operations mode over the period 2013-2016 for high-latitude regions during December and January.

Text S2.

60 Calculating cloud occurrences involves statistical uncertainties. All the different statistical parameters calculated for this study are shown in Figure S5. They are calculated from each grid cell ($2^\circ \times 2^\circ$) averaged

over 1 week of measurements, resulting in a study area represented by approximately 680,000 points. The medians are not the same as those shown in the main figures; they are not calculated on the same time scales (monthly/annual). For the whole area, the average cloud occurrence is 55.7%, with a standard deviation of 25.4%. The standard deviation obtained (Gaussian distribution hypothesis) for the different regions of the study area varies between 26 and 32%. In this study, the Wilcoxon-Mann-Whitney and Kolmogorov-Smirnov tests (not shown here) are used to evaluate the statistical significance of the observed disparities between regional cloud occurrences. Statistical analysis indicates that the cloud occurrence distribution does not align with the normal hypothesis and that these distributions vary significantly between the different regions that are studied. Therefore, we choose to use the median cloud occurrence and the Median Absolute Deviation (MAD), which are more robust parameters to characterize non-Gaussian samples.

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