

# Improvements in cloud climatology with CALIPSO and CloudSat missions

dr Andrzej Z. Kotarba (D) http://orcid.org/0000-0001-7982-1992

Centrum Badań Kosmicznych PAN akotarba@cbk.waw.pl

IGF UW – Warszawa – 2021/05/28



#### Outline

CLOUDSAT and CALIPSO two-slide intro

#### 1. CALIPSO helps MODIS

 How the CALIPSO profiles collocated with MODIS cloud detections allowed for a calibration of MODIS cloud climatology

#### 2. CALIPSO validates SYNOP

 How the CALIPSO data on cirrus validated the only pre-satellite climate records on cirrus, originating from surface-based, manual observations

#### 3. CALIPSO joins CloudSat

 How the unique, joint CloudSat-CALIPSO vertically-resoled cloud climatology was impacted by a mission-specific sampling scheme

#### • **CALIPSO**: <u>Cloud-Aerosol Lidar and Infrared Pathfinder Satellite Observation</u>

- CALIOP = Cloud-Aerosol Lidar with Orthogonal Polarisation; 0.532  $\mu$ m and 1.064  $\mu$ m
- Footprint: 100 m, sampling every 333 m (also averaged into 1 km and 5 km)
- Vertical resolution: 30 m (below 8.2 km), and 60 m (8.2-20.2 km); up to 40 km
- Sensitive to optically thin clouds (COD <0.01), signal totally attenuated at COD ~5</li>

#### CloudSat

- Cloud Profiling Radar (CPR) at 94 GHz
- Footprint: 1.4 km (cross-track) × 1.7 km (along-track); vertical resolution 480 m, up to 25 km
- Sensitive to optically thick clouds / hydrometeors

#### Orbited in a close orbital formation 2006-2011

- Launched together in April 2006
- Joined A-Train constellation
- Temporal separation 15 sec
- 60 sec after Aqua (MODIS)

705 km orbit sun-synchronous EQT 13:30 LST

#### **CALIPSO and CloudSat**

 $\rightarrow$  Complementary mission of lidar and radar



- MODIS: Moderate Resolution Imaging Spectroradiometer \_\_\_\_\_
  - state-of-the-art cloud imager on board Terra (1999+) and Aqua (2002+) polar orbiting satellites
  - 36 spectra bands (0.4  $\mu m$  14.4  $\mu m$  ), some dedicated for clouds
  - very stable in terms of radiometry, and orbit = best qality data for climate studies
  - sensitive to clouds with optical depth >0.4 (Ackerman et al., 2008)
- Why MODIS needs help?
  - MODIS does not inform on cloud amount, but only report four classes for cloud detection: confident clear, probably clear, probably cloudy, confident cloudy.
    - 87% for MODIS Collection 005 (Holz et al. 2008)
    - 77.8% for Collection 006 (Wang et al. 2016)
    - 86.7% for Collection 061 (Kotarba 2020)
  - Big question is: How those classes translate into a quantitative measure?
    What is the fractional cloud cover that should be assigned to those classes?

Assumption by NASA is

- confident clear, probably clear = fractional cloud cover of 0%
- probably cloudy, confident cloudy = fractional cloud cover of 100%

• NASA assumption can be validated by looking inside a MODIS 1 km FOV...



e.g. ASTER 30 m vs MODIS 1 km (Kotarba 2010; 10.1029/2009JD013520), but it works only on a limited scale.

• ... or alternatively use CALIPSO data to test.



- Match CALIPSO profiles with MODIS IFOVS
- pros: collocated observations MODIS Aqua + CALIPSO (A-Train constellation)
- cons: not exactly the same FOV (but it's better than pure guessing)

#### Experiment

- MODIS MYD35 Collection 061 + CALIPSO CloudLayer L2 1 km ver. 4.20
- January and July 2005, 33 793 648 MODIS-CALIPSO pairs, dt = 81 s (avr.)
- Assumption: cloud detected by CALIOP fills whole MODIS IFOV (100% cloudy)

**Full results in:** Kotarba A.Z., (2020) *Calibration of global MODIS cloud amount using CALIOP cloud profiles*. Atmospheric Measurement Techniques, 13, 4995-5012, doi:10.5194/amt-13-4995-2020.

#### $\rightarrow$ Imager vs lidar's profiles

Sun-Mack et al. (2007)



 $\rightarrow$  The actual cloud fraction for CM classes is...



 $\rightarrow$  The actual cloud fraction for CM classes is...



 $\rightarrow$  The actual cloud fraction for CM classes is...

Source of cloud fractions for		Cloud fractions (%) for MODIS cloud mask class					
cloud mask classes		(class frequency, % of <i>n</i> )					
		Confident clear (28.9%)	Probably clear (7.5 %)	Probably cloudy (5.8%)	Confident cloudy (57.8 %)		
Operational	Day+night	0.0	0.0	100.0	100.0		
This study	Day+night	21.5	27.7	66.6	94.7		
	Day only	12.7	28.4	58.4	94.7		
	Night only	29.5	27.1	70.7	94.7		

#### $\rightarrow$ Fractions for algorithm paths...

	MODIS Cloud Mask Test Layout for a Given Processing Path									
	Daytime Ocean	Nighttime Ocean	Daytime Land	Nighttime Land	Polar Day (snow)	Polar Night (snow)	Coastline Day	Coastline Night	Desert Day	Desert Night
BT <sub>11</sub> (bit 13)	1	1								
$BT_{139}$ (bit 14)	1	1	1	1	1	1	1	1	1	1
$BT_{67}$ (bit 15)	1	1	1	1	1	1	1	1	1	1
$R_{1,38}$ (bit 16)	1		1		1		1		1	
$BT_{37} - BT_{12}$ (bit 17)				1		✓				1
BT <sub>8-11</sub> & BT <sub>11-12</sub> (bit 18)	1	1	1	1			1	1	1	1
$BT_{37} - BT_{11}$ (bit 19)	1	1	1	1	1	✓	1	1	1	1
R 66 or R 87 (bit 20)	1		1				1			
$R_{87}/R_{0.66}$ (bit 21)	1		1							
$R_{935}/R_{87}$ (bit 22)	1		1		1		1			
$BT_{37} - BT_{39}$ (bit 23)	1		1				1		1	
Temporal consistency (bit 25)	1	1								1
Spatial variability (bit 25)	1	1								

Table 3. MODIS Cloud Mask Tests Executed ( $\checkmark$ ) for a Given Processing Path

Ackerman et al. 1998

- 13+ "sub-algorithms" = possible inhomogeneity in cloud detection

 $\rightarrow$  Fractions for algorithm paths...

Cloud masking algorithm path				CALIOP-based cloud fractions (%) for MODIS cloud mask class				
				Confident clear	Probably clear	Probably cloudy	Confident cloudy	
Day (47.2)	Snow-covered (5.5)	Land Desert Coast Ocean	(0.2) (3.9) (0.2) (1.1)	13.8 12.6 15.3 20.5	67.0 32.6 55.5 76.3	56.0 71.8 61.8 69.7	97.6 96.6 93.8 88.6	
	Snow-free (41.7)	Land Desert Coast Ocean	(6.7) (3.4) (1.6) (30.1)	15.6 9.1 19.0 10.5	32.3 19.1 33.8 28.4	63.9 45.5 59.8 54.5	93.4 90.0 93.0 95.2	
Night (52.8)	Snow-covered (15.8)	Land Desert Coast Ocean	(2.6) (4.7) (0.9) (7.6)	31.4 34.3 29.8 49.7	65.0 65.3 60.8 73.7	80.9 75.9 75.0 82.5	93.9 86.4 93.7 96.8	
	Snow-free (37.0)	Land Desert Coast Ocean	(5.4) (2.6) (0.9) (28.1)	8.0 8.2 10.9 22.9	25.6 23.5 23.0 22.4	68.5 55.8 60.9 61.8	97.7 95.4 96.4 94.6	

0%

100%



Figure 6. CALIOP-based cloud fraction for MODIS cloud mask classes for the "daytime, snow-free land" algorithm path and corresponding histograms (red vertical line indicates the mean value).



Figure 7. CALIOP-based cloud fraction for MODIS cloud mask classes for the "daytime, snow-free ocean" algorithm path and corresponding histograms (red vertical line indicates the mean value).



Figure 5. CALIOP-based cloud fraction for MODIS cloud mask classes for the "nighttime snow-free land" algorithm path and corresponding histograms (red vertical line indicates the mean value).



Figure 4. CALIOP-based cloud fraction for MODIS cloud mask classes for the "nighttime, snow(ice)-covered ocean" algorithm path and corresponding histograms (red vertical line indicates the mean value).

→ Practical implications: Calibration of MODIS climatology



#### CALIPSO helps MODIS → Conclusions

- Method for deriving empirical cloud fraction for thematically cloud detection classes; applicable to MODIS, AVHRR, VIIRS, and geostationary imagers.
- MODIS ST assumption is inaccurate; the actual cloud fractions for MODIS are: 21.5 %, 27.7 %, 66.6 %, and 94.7 %, instead of 0%, 0%, 100%, 100%.
- Cloud fractions vary among algorithm paths, and regionally within a single path (region specific cloud detection errors).
- Calibrated climatology indicate over/under estimation of as much as 30% in polar regions, and "aerosol regions".

- Surface-based detection of cirrus (SYNOP)
  - Cirrus, Cirostratus, Cirrocumulus as defined by WMO hereinafter "cirrus"
  - Detected visually, by a human observers at meteo stations over land (and oceans)
  - The only pre-satellite data on cirrus, the longest existing climate records on cirrus
    United States, 1948-1994, and the former Soviet Union, 1936-1990 (Sun et al. 2001), Canada, 1953–2003 (Milewska 2008), China, 1971-1996 (Endo and Yasunari 2006), the Arctic, 1954-2008 (Eastman and Warren 2010), the northern Chilean coast, 1969-2013 (Muñoz et al. 2016), the north–east of Spain, 1910-2006 (Curto et al. 2009), or Poland, 1971-2000 (Filipiak and Miętus 2009).
  - Challenging geometry: cloud overlap; unknown sensitivity of human eye to cirrus

#### Experiment

- Quality-checked SYNOP FM-12 reports for 2006-2020, globally
- CALIPSO: CloudLayer L2 5 km ver. 4.20 (the most sensitive product for cirrus)
- Confusion matrix for binary classification, and related measures of agreement

**Full results in:** [1] Kotarba A.Z., Nguyen Huu, Ż., *Accuracy of visual cirrus detection by a surface-based human observers. (about to be submitted);* [2] Nguyen Huu, Ż., Kotarba, A.Z., (2021) Reliability of visual detections of cirrus over Poland. Theoretical and Applied Climatology, 144, 1–11, doi:10.1007/s00704-020-03494-9.

#### $\rightarrow$ Data base of paired observations



• dxy = 11 km (avr.), VV=33 km (avr.)

 $\rightarrow$  Field of view issue (inconsistency)



 $\rightarrow$  Field of view issue (inconsistency) – correction factor

• How frequently CAIPSO passes over a station, and report no cirrus just because of a ground track misalignment?

Simulation

- MODIS data: Aqua, full day of observations, every 10 days in 2017 = 10,368 MODIS granules
- International Satellite Cloud Climatology Project (ISCCP) definition of cirrus: cloud with optical thickness less than 23, and top pressure less than 440 hPa
- virtual meteo station every 100 km along the MODIS ground track
- simulated CALIPSO pass 11 km to the station (i.e. average dxy in SYNOP-CALIPSO database)
- SYNOP-like cirrus reported within 33 km buffer (i.e. average visibility range in our database)



100 km @ nadir

 $\rightarrow$  Field of view issue (inconsistency) – correction factor

• How frequently CAIPSO passes over a station, and report no cirrus just because of a ground track misalignment?

#### • Simlulation

- MODIS data: Aqua, full day of observations, every 10 days in 2017 = 10,368 MODIS granules
- International Satellite Cloud Climatology Project (ISCCP) definition of cirrus: cloud with optical thickness less than 23, and top pressure less than 440 hPa
- virtual meteo station every 100 km along the MODIS ground track
- simulated CALIPSO pass 11 km to the station (i.e. average dxy in SYNOP-CALIPSO database)
- SYNOP-like cirrus reported within 33 km buffer (i.e. average visibility range in our database)

#### • Results

- 13% of observations: cirrus missed by CALIPSO because of ground track mislocation
- 32% of observations: cirrus presence confirmed by both techniques
- 55% of observations: cirrus absence confirmed by both techniques
- Correction factor: **19%** of FP  $\rightarrow$  TP, TN  $\rightarrow$  FN
- + Sensitivity study (now the corr. factor ipmacts the results)

 $\rightarrow$  Overall agreement

	Conditions		Accu	ıracy m	easures		
		POD	FAR	OA	F-score	к	
method's	All, day	48%	22%	60%	0,58	0,24	
performance	All, night	28%	12%	53%	0,40	0,14	
	Perfect, day	<mark>67%</mark>	24%	72%	0,69	0,43	observer's
	Perfect, night	35%	13%	61%	0,47	0,22	performance

- Probability of cirrus detection (POD), false alarm rate (FAR), overal accuracy (OA)
- F-score + Cohen's Kappa (evaluation of binary clasification; range 0 to 1)

 $\rightarrow$  Perfect conditions: cirrus optical depth



- Very strong dependance, both day-time and night-time
- Interestingly: sub-visual cirrus ( $\tau_{cir}$ <0.03) also detected by human observers

 $\rightarrow$  Perfect conditions: aerosol misclassified as cirrus?



- Possibly during the night time, but relations is weak
- Day-time conditions are inconclusive

 $\rightarrow$  Perfect conditions: any hope from lunar illumination?



• Probability of detction slightly higher when lunar pahse >50%, no impact on FAR and OA

 $\rightarrow$  Real conditions: clouds at middle and/or low levels



- Performance systematically decreases as the cloud fraction increase
- Rapid drop in POD after 60% day time, and 25% night time.
- Kappa always very low (almost random agreement between SYNOP and CALIPSO)

 $\rightarrow$  Real conditions: clouds at middle level only



 $\rightarrow$  Correction factor – sensitivity study



- Corr. factor = 19% is close to the best of what can be achieved with SYNOP
- Even with any other corr. factor (0-100%) the performance still would be moderate to low

 $\rightarrow$  Conclusions

- First quantitative assessment of visual detection of cirrus; method that allows for matching SYNOP and CALIPSO transects + method's uncertainty analysis.
- SYNOP moderately reliable (PoO>60%) for cirrus only daytime, only during perfect condition or under real conditions but with few middle/low-level clouds
- In other cases detections unreliable (PoD<50%) → agreement with CALIPSO can be purely random (very low Kappa coincident, <0.2).</li>
- Lunar illumination is not much helpful for cirrus detection.
- The results can be also a benchmark for camera-based detections (does your algorithm perform better than a human observer?)

- The only avaliable lidar—radar cloud profile data, globally (2006-2011) (Follow on: EarthCARE 2021+, or Aerosol Cloud Convection and Precipitation ?)
- Complementary observations: CALIOP (thin clouds) + CloudSat (thick clouds)
- Profiling instruments + 16-day revisit = 22/23 obs. per year
- Uncertanity of climate data resulting from the infrequent revisit...?

#### Experiment

- 2B-GEOPROF-LIDAR (ver. P2\_R05), Layer Base and Layer Top altitude (m)
- Epoch 00 to Epoch 04 (2 June 2006 17 April 2011), dt = 10-15 s
- Mean cloud amount at 40 levels (480 m) + bootstrap confidence interval at: 4 spatial resolutions, 4 confidence levels, and 3 time scales.

**Full results in:** Kotarba, A.Z.; Solecki, M. (2021) Uncertainty Assessment of the Vertically-Resolved Cloud Amount for Joint CloudSat–CALIPSO Radar–Lidar Observations. Remote Sensing, 13, 807. doi: 10.3390/rs13040807.

 $\rightarrow$  Orbit and sampling



#### $\rightarrow$ 3D climatology + uncertanity analysis



• Only the troposphere has been considered in this study

#### $\rightarrow$ Average width of confidence interval

**Table 2.** Average width of the Confidence Interval (CI) for assumed Confidence Levels (CL), and grid box sizes.

Confidence		ce	Width of Conf	idence Interval (%)	
L	evel (C	L) 0.85	0.90	0.95	0.99
			Ann	ual mean	
	1.0°	6.10	6.96	8.27	10.81
	2.5°	3.42	3.91	4.66	6.12
	5.0°	2.19	2.50	2.98	3.91
	10.0°	1.38	1.58	1.88	2.47
			Seasonal 1	nean (autumn)	
	1.0°	10.71	12.19	14.44	18.67
	2.5°	6.20	7.08	8.43	11.04
	5.0°	3.98	4.55	5.42	7.12
	10.0°	2.53	2.89	3.45	4.53
			Monthly m	ean (September)	
	1.0°	16.48	18.66	21.87	27.59
	2.5°	10.14	11.56	13.72	17.83
	5.0°	6.61	7.55	8.98	11.75
	10.0°	4.26	4.86	5.79	7.60

5-year data (2006-2011)

 $\rightarrow$  Average width of confidence interval



- **4× increase** ← resolution of a grid increases from 10° to 1°
- 3× increase ← number of months considered decrease from 12 (annual) to 1 (monthly)
- 2× increase ← confidence level increases from 0.85 to 0.99



 $\rightarrow$  Vertical structure of confidence interval





5-year annual / 2.5 deg / conf. lev. 0.95

#### $\rightarrow$ Expected accuracy

• Global Climate Observing System (GCOS) (WMO 2011)

1% (optically thin clouds) to 5% (optically thick clouds)

+ cloud data to be available globally, every three hours, at a spatial resolution of 50 km (0.5° at the equator)

#### • US National Institute of Standards and Technology (NIST) (Ohring et al., 2005)

1% accuracy for global mean cloud cover

 $\rightarrow$  Expected accuracy (1%-5%)

Fraction of volumes in the atmospheric column (%)



1%  $\rightarrow$  6.5% of volumes at 1° and 2.5° 5%  $\rightarrow$  22.5% of volumes at 1° 48.9% of volumes at 2.5°



 $\rightarrow$  Detectable change in cloud structure  $\rightarrow$  Chepfer et al. (2014)



- "Current climate" minus "+4K climate", two models (CanAM4, HadGEM2)
- Simulated lidar data for model (lidar only)
- + real variability of CALIPSO (GOCCP 2006-2012)

#### $\rightarrow$ What matters most?

**Table 3.** Partial contribution of variables to the model's determination coefficient ( $R^2$ ). Partial coefficients are not scaled, i.e., they sum up to the overall value of  $R^2$ . Results refer to cloud amount analyzed in the annual timeframe.

	Partial Contribution to R2				
Grid size:	1°	2.5°	5°	<b>10°</b>	
Model's overall R <sup>2</sup> :	65.3	87.6	89.8	91.1	
mean cloud amount	14.7	20.2	21.0	21.4	
std. dev. of cloud amount	30.9	37.5	37.5	36.4	
Geography					
latitude	< 0.1	< 0.1	< 0.1	< 0.1	
longitude	0.1	0.1	0.1	0.1	
altitude	1.9	2.6	2.0	2.3	
Statistical					
no. of observations	6.6	8.8	10.5	9.5	
Confidence Level	11.1	18.4	18.7	21.4	

- Some variable are not 100% independent...
- Still: a general insight into factors that controll the conf. interval width

#### CALIPSO joins CloudSat → Conclusions

- First uncertainty assessment for the joint CloudSat-CALIPSO vertically-resolved cloud amounts: confidence intervals at common conf. levels and grid sizes.
- Quantitative information on how the conf. interval width is determined by: grid size, time frame for data averaging, and confidence level.
- Not possible to get the cloud amounts at high (1-5%) accuracy, while maintaining a high spatio-temporal resolution (it still may be possible locally – use our 3D data to test it for your area of interest!)
- CloudSat-CALIPSO-like configuration will most likely allow for detecting trends in optically thin clouds globally, and thick clouds in tropics and polar regions.

- [CALIOP helps MODIS] Use CALIOP detections to calibrate MODIS (AVHRR, VIIRS) cloud data → you will avoid ±30% error in cloud amount regionally.
- [CALIOP vlidates SYNOP] Use surface-based data on cirrus frequency with extreme caution → only trust the data taken under perfect conditions, and daytime.
- [CALIOP joins CloudSat] Consider uncertanity budget for joint CloudSat-CALIPSO cloud climatology (may by very large for you 3D location) → do not expect high accuracy at fine spatio-temporal resolution.

CBK PAN is open for collaboration on MODIS/CALIPSO/CloudSat/and other imagers/souders  $\rightarrow$  MSc/PhD interns are welcome too!

#### Thank you for attention!

#### Acknowledgement

- Funded by NCN (LiRaC UMO-2017/25/B/ST10/01787)
- Funded by CBK PAN (statutory theme 2020.10)
- Funded by **PLGrid** (high-performance computing resources)

