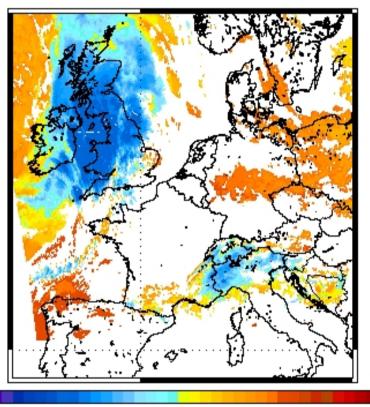
# **Satellite-based VIS/IR multispectral screening of precipitating clouds:** A case study during summer at mid-latitudes $(\mathbf{i})$



# **1. INTRODUCTION**

- $\diamond$  The detection of precipitating clouds from geostationary sensors provides a characterization on time scales consistent with the nature and development of precipitating systems and oriented to potential operational applications.
- $\diamond$  The aim is to implement a precipitating cloud (PC) detection methodology based on MSG spectral channels and rain rates from the NIMROD radar network and to evaluate its performances against NIMROD data and other satellite based PC detection techniques.

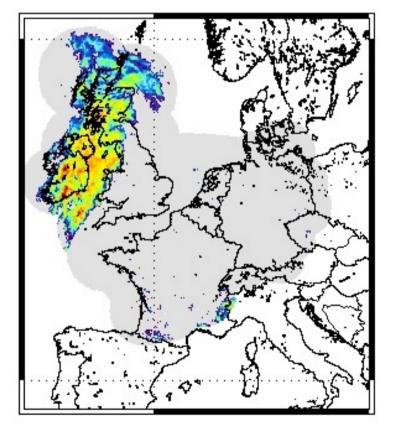
## 2. DATA SETS

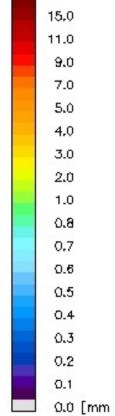


# **MSG-SEVIRI DATA**

17/06/2009 0630 UTC BT at 10.8 µm

- $\diamond$  Data from summer 2009 (JJA)  $\diamond$  Channels: 0.6, 1.6 (daytime), 3.9 (nighttime), 6.2, 8.7, 10.8, and 12.0 µm
- $\diamond$  Cloudy pixel selection: CTTH and CEFFECT (PGE03) from SAF NWC (https://www.nwcsaf.org)
- ♦ Correction for parallax error: CTTH and routine elaborated by the Convection Working Group (http://www.convection-wg.org)





**NIMROD c-band radar** network: rain rate product (http://badc.nerc.ac.uk)

17/06/2009 0630 UTC

- $\diamond$  North-West Europe composite: 5 km spatial resolution, 15 min refresh time
- $\diamond$  A matching method based on the distance between MSG and radar pixels was implemented to determine if a MSG pixel is covered by PC or not (rain rate > 0.1 mm  $h^{-1}$ )

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### **3. METHODOLOGY**

(Thies et al., Atmos. Chem. Phys., 2008; Thies et al., Meteorol. Appl., 2008

$$PI(x_1,...,x_n) = \frac{N_p(x_1,...,x_n)}{N_p(x_1,...,x_n) + N_{np}(x_1,...,x_n)}$$

Look-up tables of Probability Index (PI)  $\diamond$  July and August 2009 data sets  $x_1, \dots, x_n$ : MSG spectral features, 0.6/1.6 (reflectances ratio, daytime), BTD(3.9-10.8) (nighttime), BT10.8, BTD(10.8-12), BTD(6.2-10.8), BTD(8.7-10.8)  $A = N_p$  and  $N_{np}$ : frequencies of PC and non-PC pixels having  $x_1, \dots, x_n$  values for the selected features

**PC detection**  $PI(x_1,...,x_n) > PI_{th}$  with threshold values determined by maximizing the Equitable Threat Score (ETS)

# 4. RESULTS and CONCLUSIONS

The daytime approach is most effective in PC detection as shown by ETS, ACC, FAR, and POFD values.

	Day	Night	Twilight	
ACC	0.88	0.76	0.84	
BIAS	1.15	1.70	1.27	
POD	0.60	0.66	0.48	
FAR	0.48	0.61	0.62	
POFD	0.08	0.21	0.11	
ETS	0.33	0.21	0.20	
НК	0.52	0.45	0.37	

June 2009 – PC detection algorithm

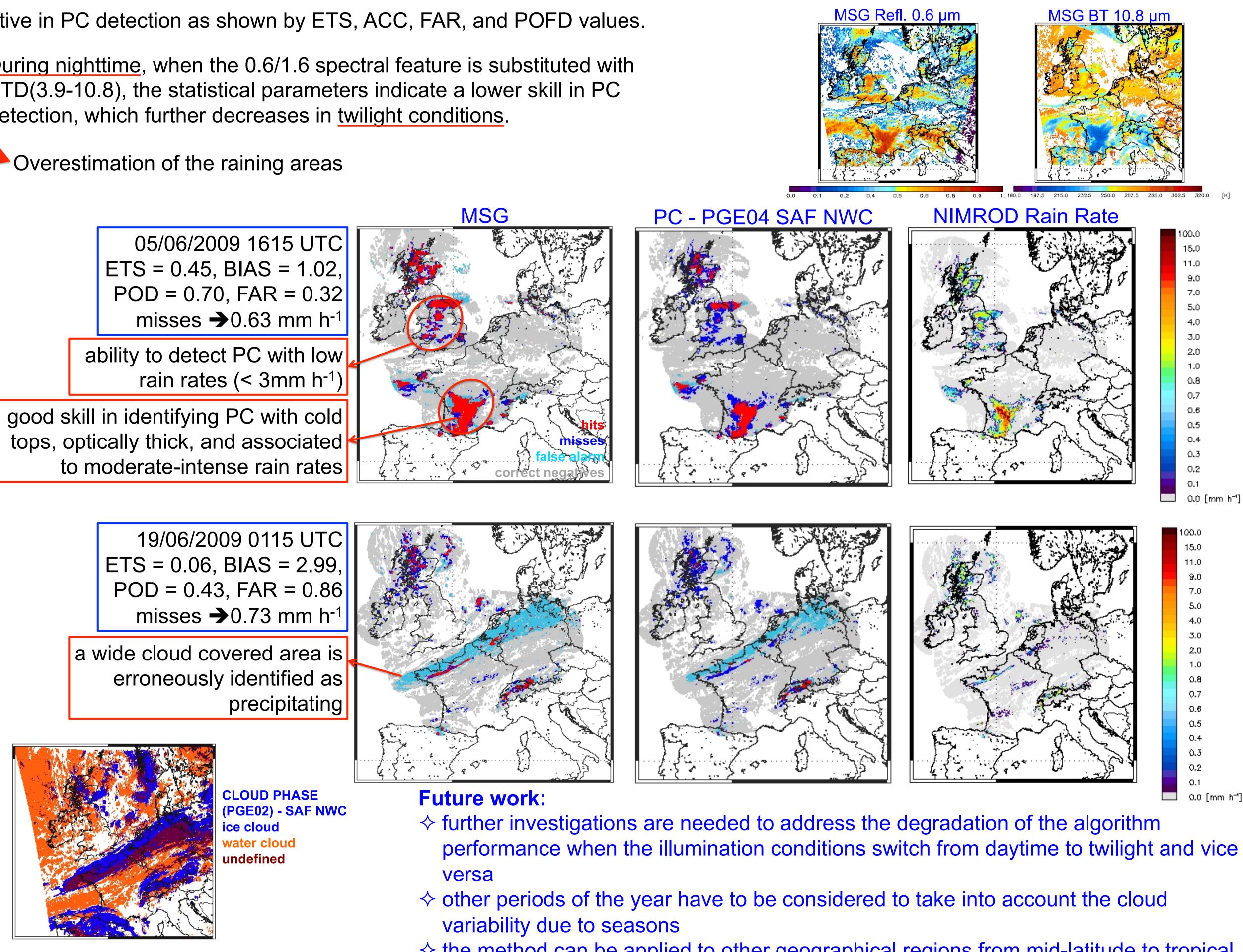
vs NIMROD rain rates

	mean	stdev
ACC	0.87	0.07
BIAS	1.45	1.89
POD	0.47	0.17
FAR	0.61	0.16
POFD	0.09	0.06
ETS	0.22	0.10
HK	0.38	0.15
PGE04	mean	stdev
ACC	0.90	0.06
BIAS	0.80	0.69
POD	0.29	0.20
FAR	0.57	0.21
POFD	0.05	0.05
ETS	0.16	0.11

June 2009 – PC detection algorithm vs Precipitating Cloud product (PGE04) from the SAF NWC

During nighttime, when the 0.6/1.6 spectral feature is substituted with BTD(3.9-10.8), the statistical parameters indicate a lower skill in PC detection, which further decreases in twilight conditions.

Overestimation of the raining areas



 $\diamond$  the method can be applied to other geographical regions from mid-latitude to tropical areas



