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Evaluation of remote-sensing snow observations for perspective of DA in NWP

/ after Maxime Quenon's Internship Report/

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thanks for Elena Shevnina

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Scientific motivation:

A gap between new advanced remote-sensing observations of snow and simple methods of snow DA in NWP

Snow Extent (SE): a binary value (yes/no).

- No statistical interpolation (OI) methods.
- No understanding of observational error.
- Difficult to combine observations from different satellites.
- In DA, we just fully trust observations.



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Practical motivation and History

No remote-sensing snow observations are assimilated into HARMONIE: the situation should be improved.

Summer 2016, Maxime Quenon, the Intern from ENSG at FMI.

The main focus of his work:

To evaluate different remote-sensing snow products for the perspective of DA in NWP.



- Results from Maxime:
Evaluation of different snow products:
 - data availability and quality
 - comparison with SYNOP data
 - comparison with model data
 - Ideas:
 - observational error for binary data - what is it? how to calculate?
 - what can be a basis to develop the statistical interpolation methods for them
- Towards improved objective analysis of snow, which combines data from different satellites



Satellite snow products

SE:

categorical value

- **MSG (METEOSAT)**
geostationary
visual band -> cloudiness problem
resolution ~ 3 km
temporal resolution: raw ~ 15 min, final ~ 1 day
- **METOP**
polar orbiting
visual band -> cloudiness problem
resolution ~ 1 km
temporal resolution: raw ~ 12 hours, final ~ 1 day

	Surface considered	No processing	Totally snow covered	Partially snow covered	No snow	Unclassified	Water
Value in the classification	0	1	2	3	4	5	

MAXIME QUENON, 2016



real value

- SYNOP
in obs points
at SYNOP times
snow depth

real value

- Harmonie
SWE
snow fraction
resolution ~ 2.5 km
every 3 hours



In situ snow data

Model snow data



Methods

- Averaging in time => data on daily basis
- Reprojecting from one grid to another using the nearest neighbor method
- Converting the binary data to the snow fraction when upcoming from the finer to coarser grid
- Converting SWE to the snow depth and back using the snow density climatology
- Converting SWE to the snow fraction and back using the ad hoc formula:

$$fraction = \frac{SWE}{SWE + 10}$$

- Converting the snow fraction to the binary SE using a threshold value of 0.5



Methods

Visual comparison

- maps

Statistics

- contingency tables
- coefficients from the contingency tables and time series of them :
- histograms and double histograms of errors (differences)

Analysis 1	Analysis 2	
	Snow	No snow
Snow	<i>a</i>	<i>b</i>
No snow	<i>c</i>	<i>d</i>



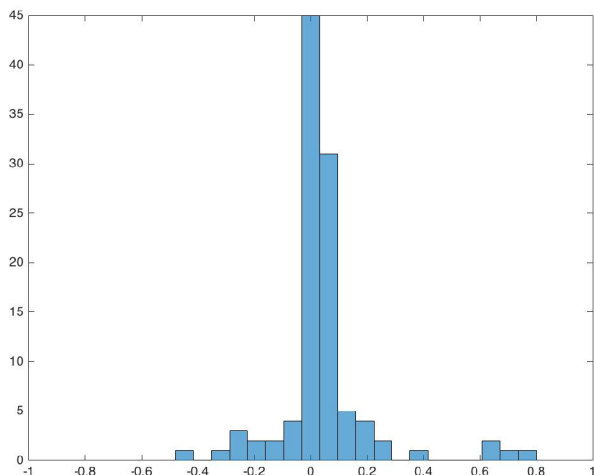
Snow data comparison

	SYNOP	MSG	METOP	Globsnow	Harmonie
SYNOP		★	★	★	-
MSG			★	-	★
METOP				-	★
Globsnow	Winter 2015-2016 Daily statistics, sampling in space Europe, Scandinavia, Finland				★
Harmonie					



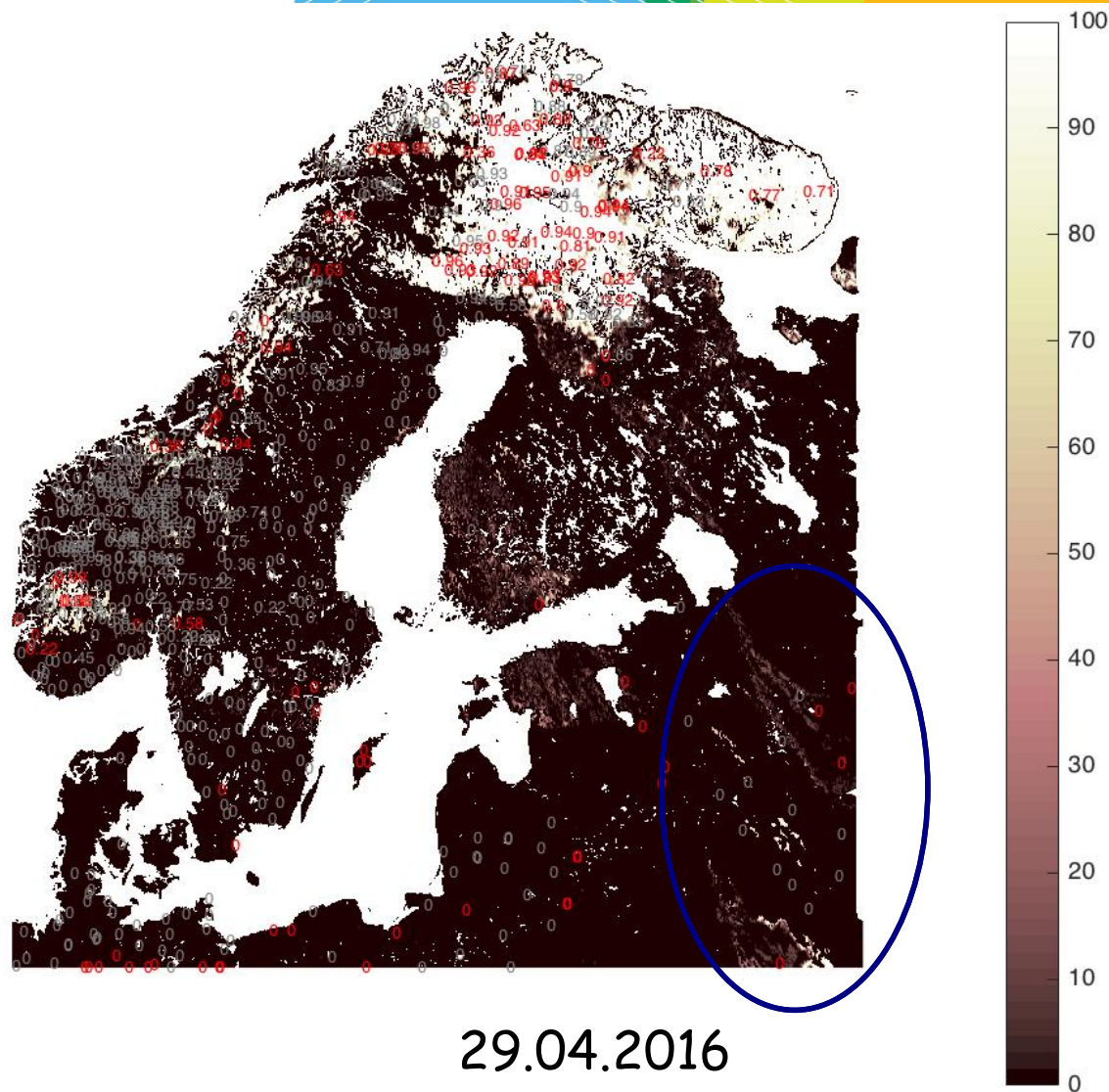
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METOP vs SYNOP



Diff=METOP-SYNOP

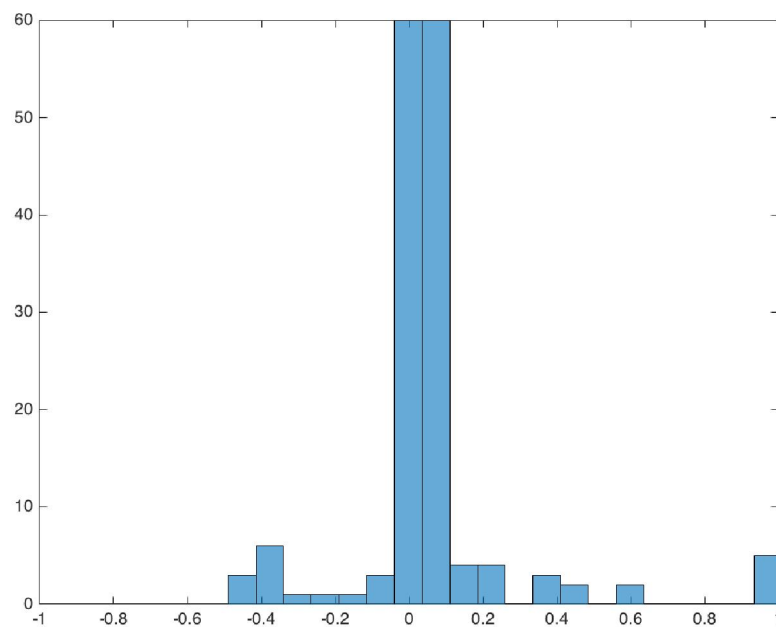
✓ "False" snow in METOP



29.04.2016



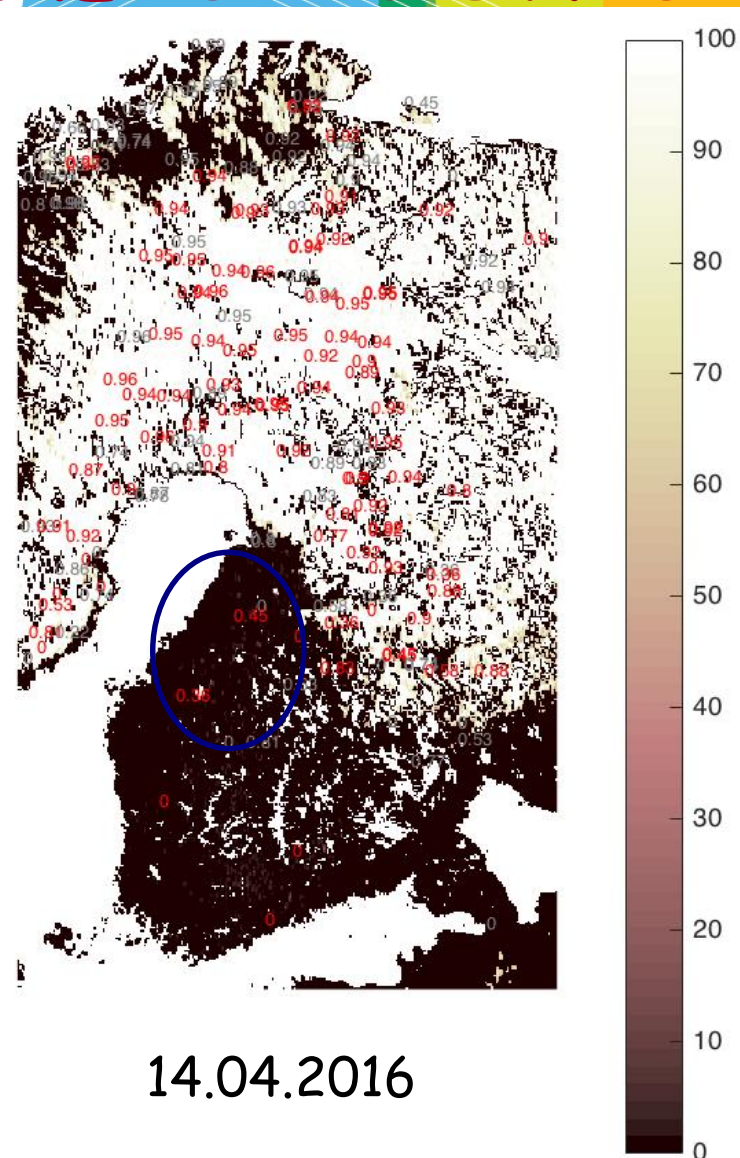
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Diff=METOP-SYNOP

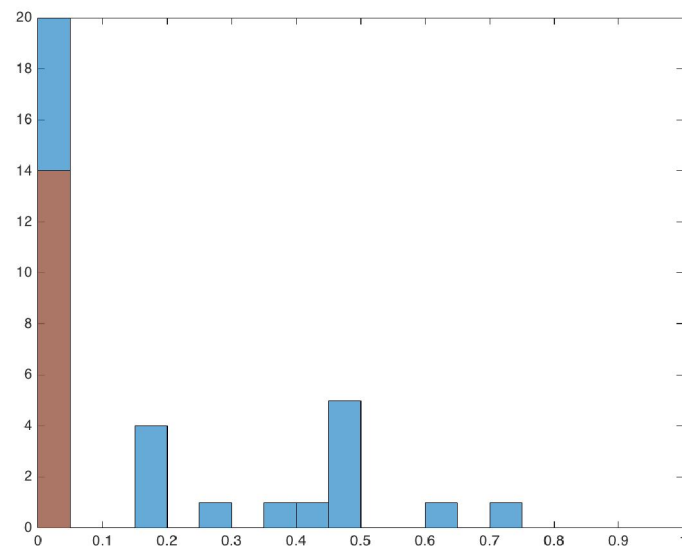
✓ "Not detected" snow in METOP

METOP vs SYNOP



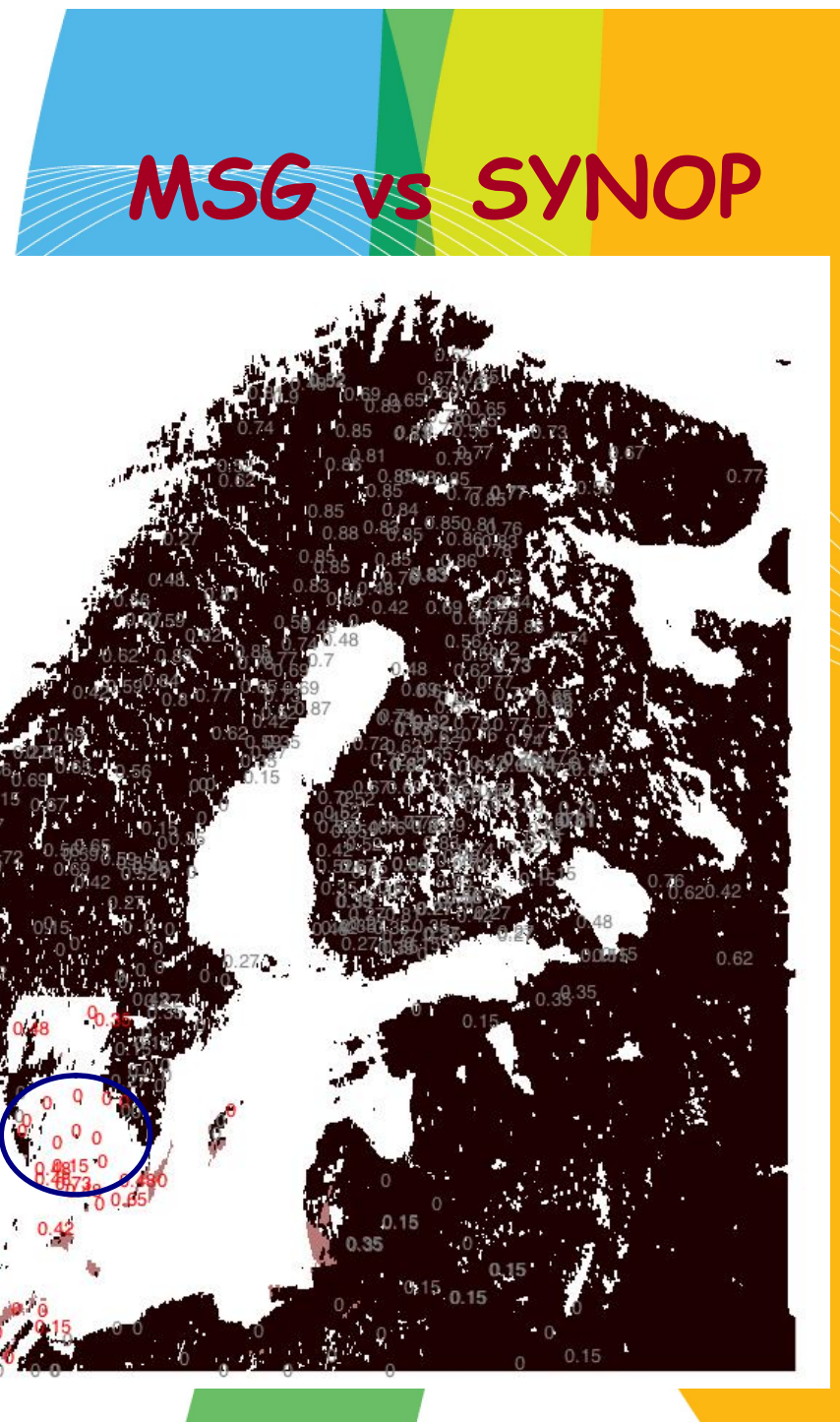


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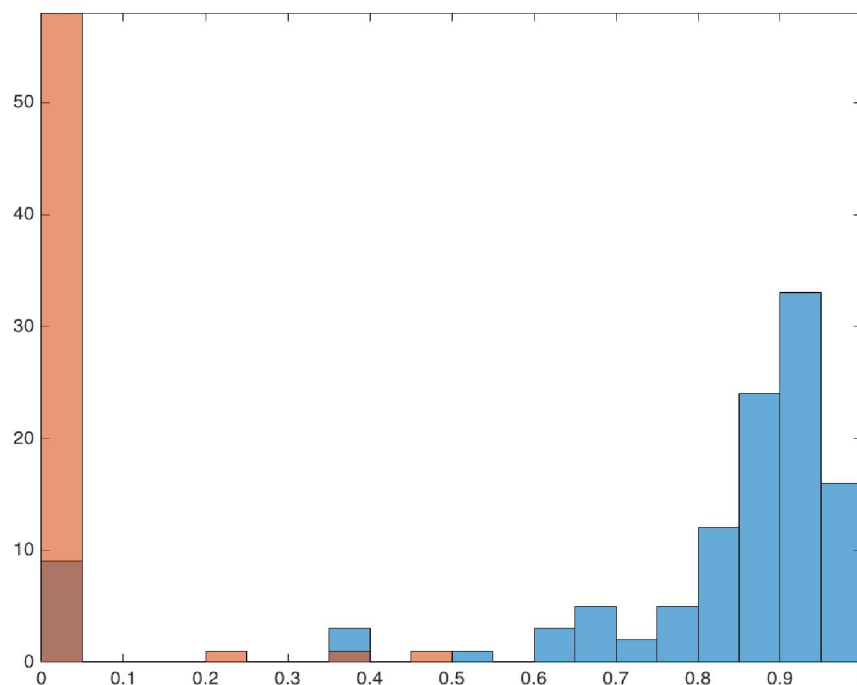
✓ "False" snow in MSG

23.11.2015





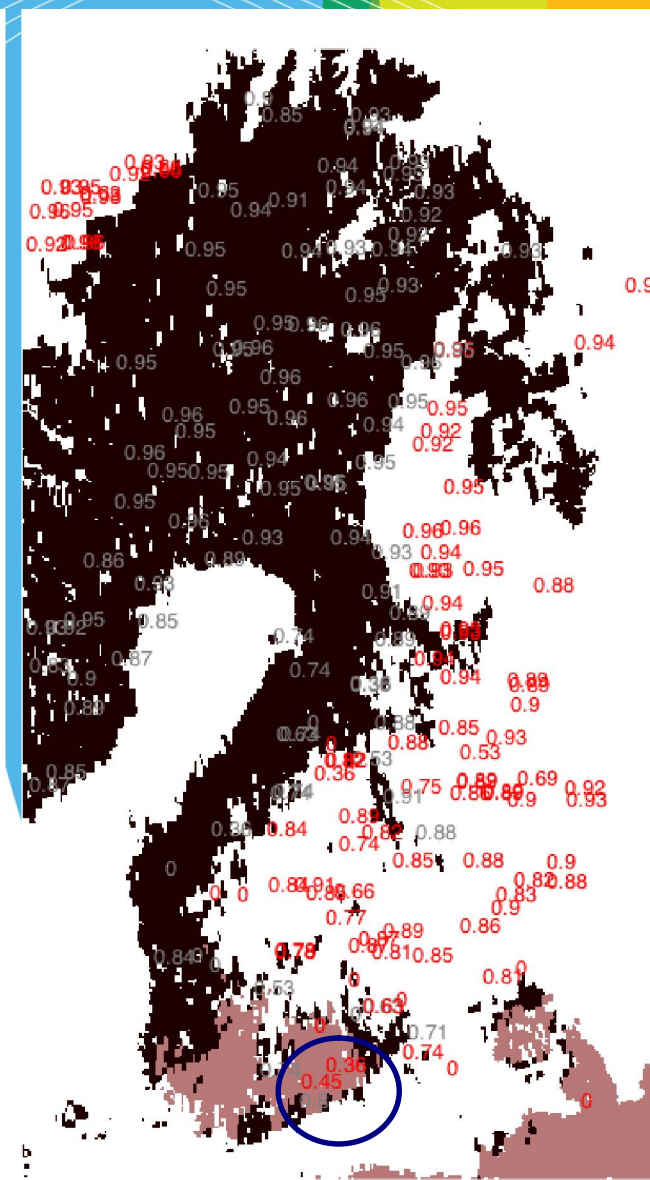
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✓ "Not detected" snow in MSG

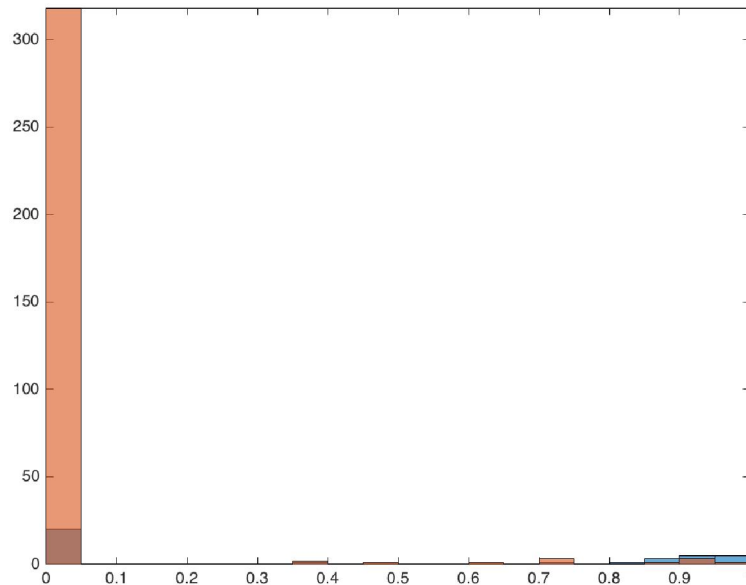
04.04.2016

MSG vs SYNOP





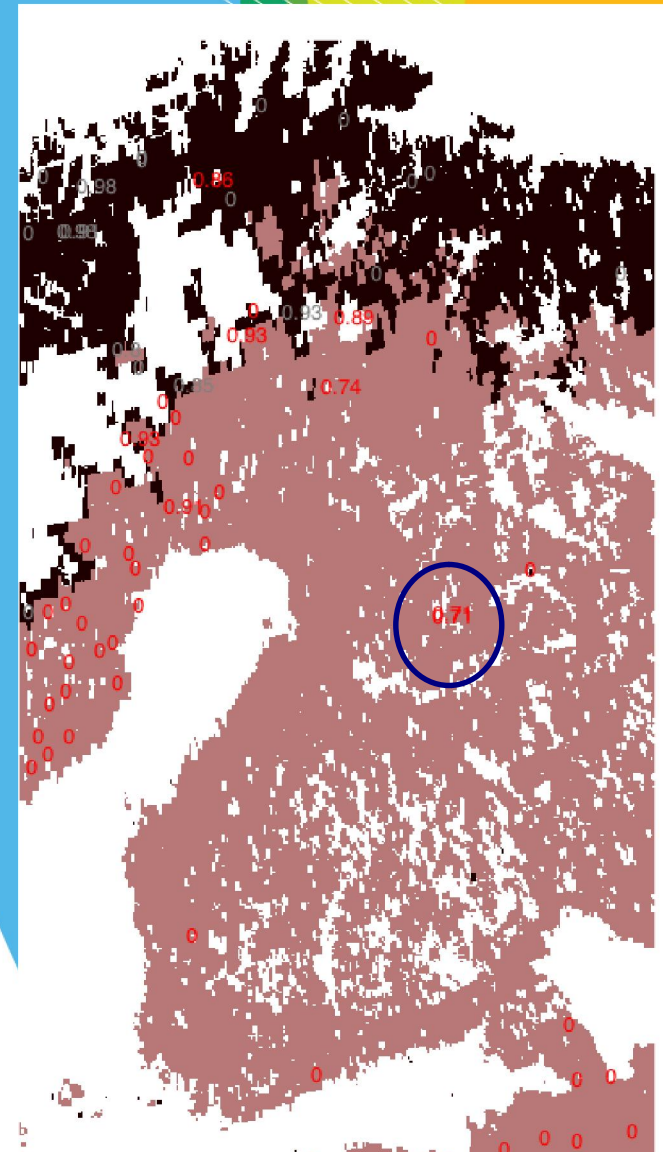
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✓ Are SYNOPs always truth?

09.05.2016

MSG vs SYNOP





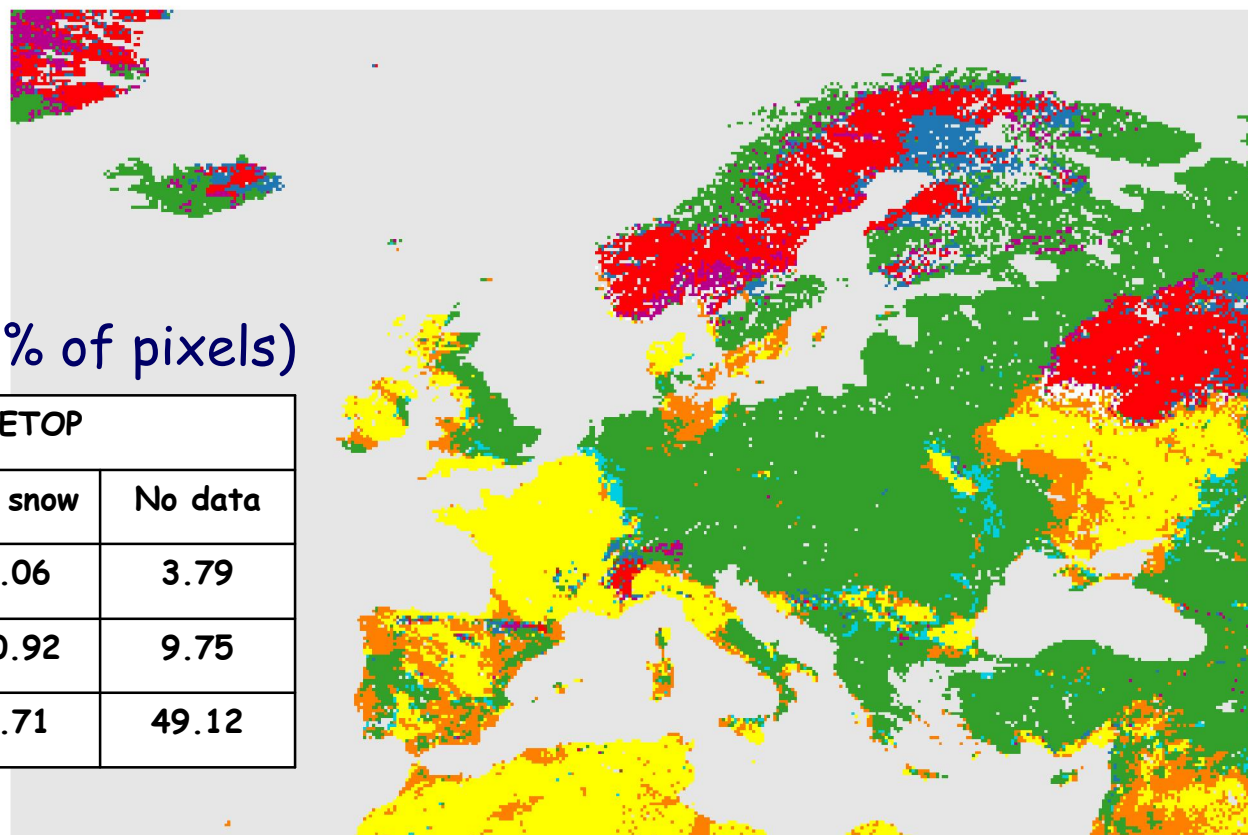
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METOP vs MSG

		METOP		
		Snow	No snow	No data
MSG	Snow	Red	Orange	Blue
	No snow	White	Yellow	Brown
	No data	Pink	Cyan	Green

Contingency table (% of pixels)

		METOP		
		Snow	No snow	No data
MSG	Snow	11.6	0.06	3.79
	No snow	0.37	20.92	9.75
	No data	2.70	1.71	49.12



15.03.2016

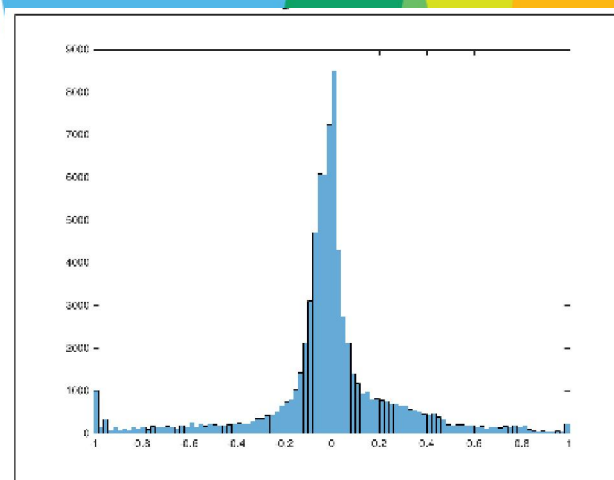
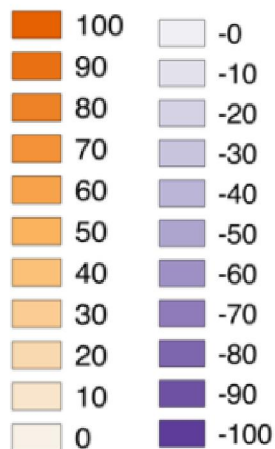
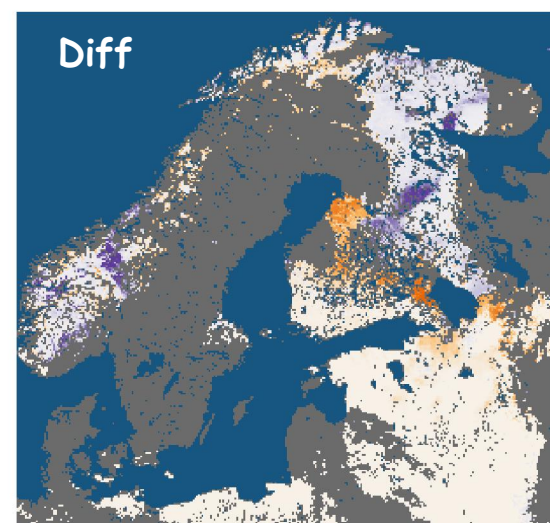
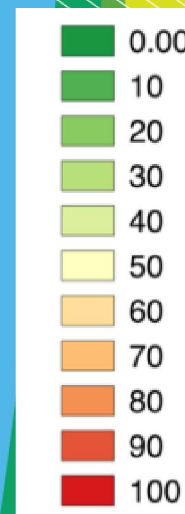
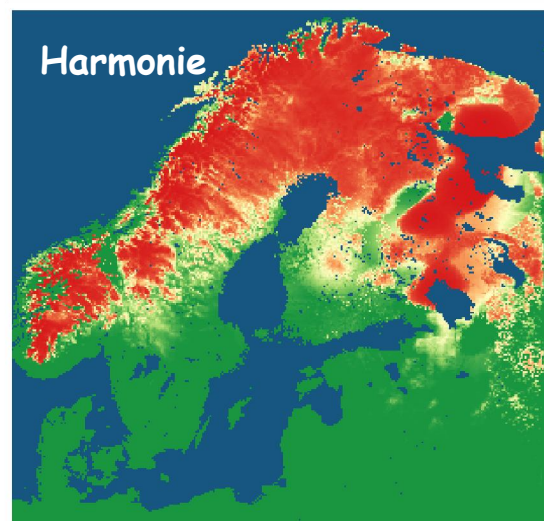
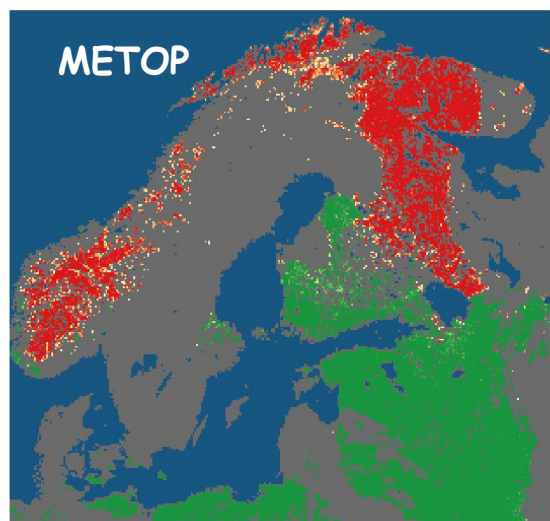


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Snow fractions

METOP vs Harmonie

16.03.2016

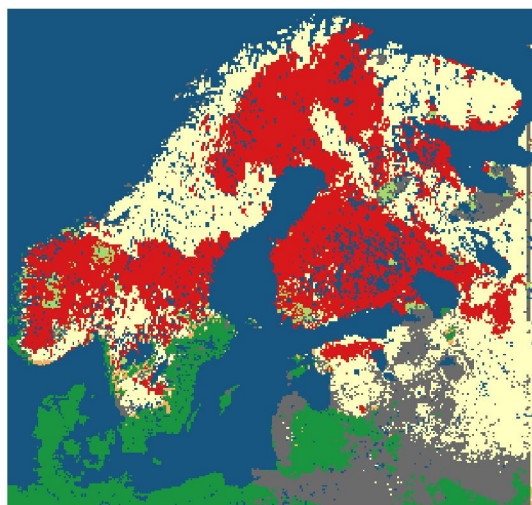
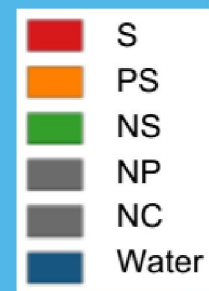
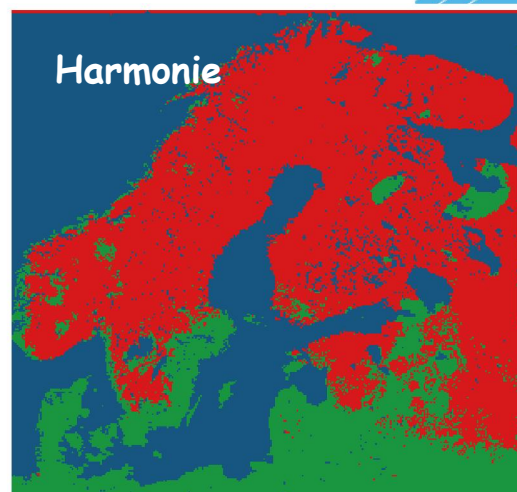
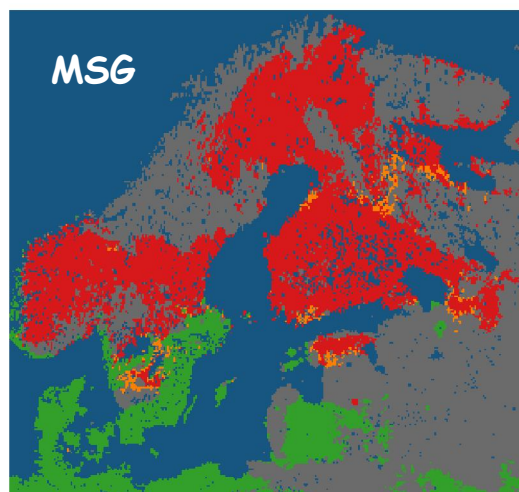


Diff=Harmonie-METOP

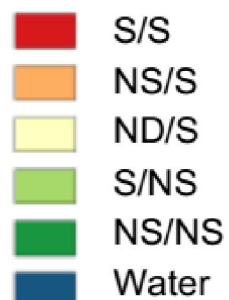


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MSG vs Harmonie



16.03.2016



Contingency table (% of pixels)

		Harmonie	
		Snow	No snow
MSG	Snow	64,76	1.47
	No snow	3.10	30.33



Conclusions from comparisons

- Overall agreement between all sources of data is good. But we look into details ... All data sources contain errors.
 - SYNOP data: representativeness errors, coding errors.
 - Both METOP and MSG data overestimate SE due to cloud contamination.
 - Both METOP and MSG may have errors of "not detected snow"
 - METOP may give the "added value" to MSG products
-
- Errors for all types of observations, as well as model errors, should be considered in NWP snow analysis
 - How to improve the analysis system?



Ideas: basis for OI

- Use Contingency tables to develop OI for SE
- Make sampling in time
- Calculate SE climatology:

$$f \begin{matrix} \nearrow \\ \searrow \end{matrix} \begin{Bmatrix} 0 \\ 1 \end{Bmatrix} \quad \begin{Bmatrix} N_1 \\ N_2 \end{Bmatrix} \quad N_1 + N_2 = M$$

$$\bar{f} = \frac{1}{M} (0 \cdot N_1 + 1 \cdot N_2)$$

$$\bar{f} = \frac{N_2}{M}$$

$$\bar{f} = p$$

- Consider the deviation:

$$f' = (f - \bar{f}) \begin{matrix} \nearrow \\ \searrow \end{matrix} \begin{Bmatrix} -p \\ 1-p \end{Bmatrix}$$



Ideas: basis for OI

- Consider the value:

$$(f'(r_1) - f'(r_2))^2$$

		$f'(r_1)$	
		$-p(r_1)$	$1-p(r_1)$
$f'(r_2)$	$-p(r_2)$	n_{11}	n_{12}
	$1-p(r_2)$	n_{21}	n_{22}

- Define a structure function:

$$b(r_1, r_2) = \frac{1}{M} [(p(r_1) - p(r_2))^2 (n_{11} + n_{22}) + \\ (p(r_2) - p(r_1) + 1)^2 n_{12} + \\ (p(r_2) - p(r_1) - 1)^2 n_{21}]$$



Ideas: basis for OI and obs error

- Define an autocorrelation function:

$$m(r_1, r_2) = \frac{1}{M} [p(r_1)p(r_2)n_{11} - (1-p(r_1))p(r_2)n_{12} - p(r_1)(1-p(r_2))n_{21} + (1-p(r_1))(1-p(r_2))n_{22}]$$

- Define an obs. error:

$$\tilde{f} = f + \delta$$

$$\overline{\delta} \approx 0 \quad \overline{\delta^2} = \frac{d_{12} + d_{21}}{M}$$

		f	
		0	1
δ	0	$0_{d_{11}}$	$-1_{d_{12}}$
	1	$1_{d_{21}}$	$0_{d_{22}}$

... etc.

- Following Gandin, 1965, calculate an obs. error and develop OI



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Thank you for your attention.

Questions?