Cloud Data Assimilation Using Observation Operator Defined by the Penalty Function. Proof of Concept

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Concept

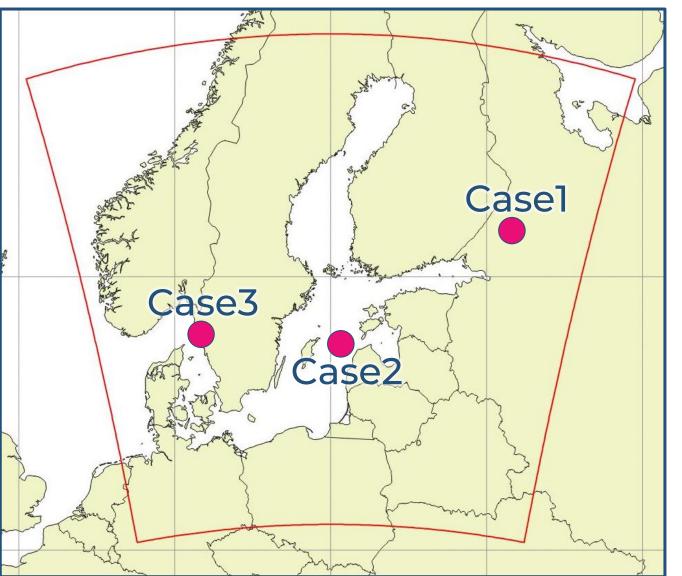
We assume the cloud layer is located between h_b - cloud base height - and h_t - cloud top height. For simplicity, we assume cloud cover is 8/8 and saturated relative humidity is 100%

It implies: RH = 1 if $h_b \le h \le h_t$

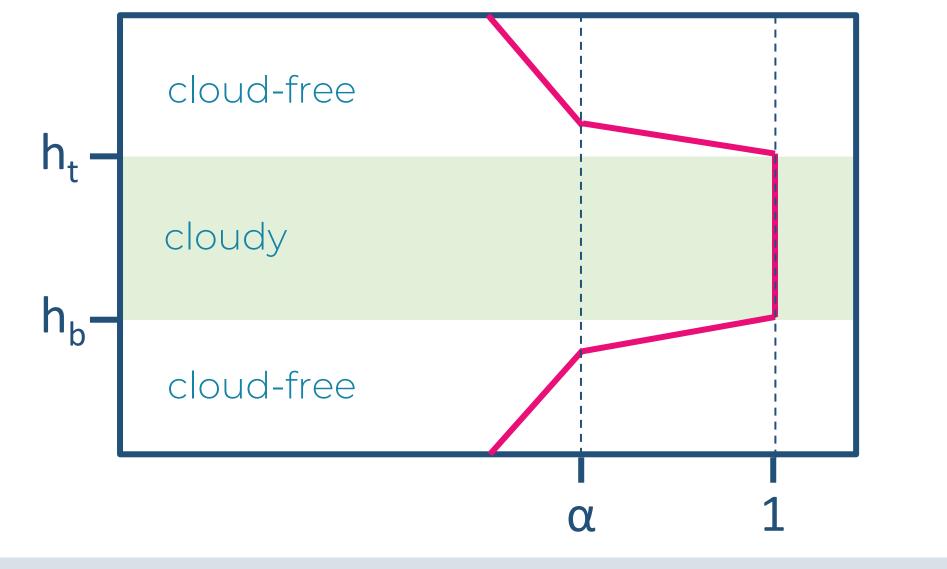
 $RH \leq \alpha < 1$ otherwise where lpha is the threshold value

Relative humidity profile

Experimental Setup



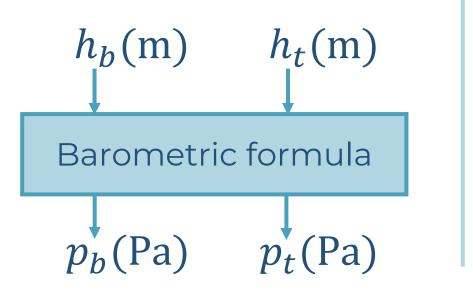
- Harmonie cy43h2.2.1 Single observation 3D-Var DA: **Domain**: HMEST25 2.5 km Grid 24/06/2023 Date: 12 to 21 UTC Time
- Casel (15 UTC) Cloud shift First guess: cloud layer at ca. 950-870 hPa Experiment: shift cloud layer to 880-800 hPa Case2 (18 UTC) - Cloud creation no low clouds First guess: Experiment: create cloud layer at 925-870 hPa

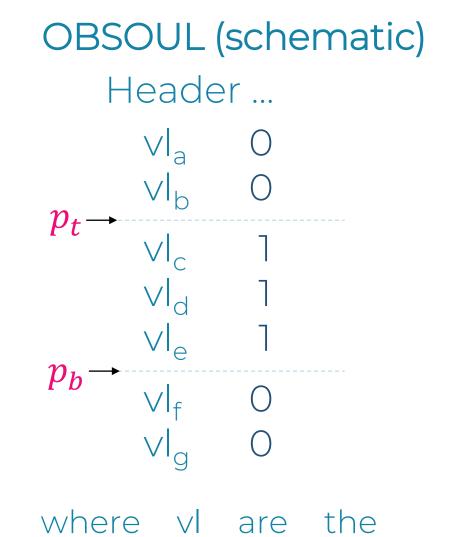


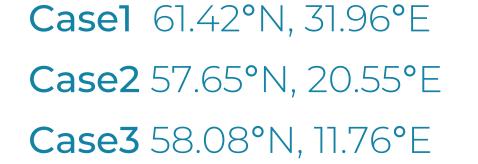
Observations

1) Ceilometer observation of cloud base altitude h_b

2) Satellite observation of cloud top altitude h_t at ceilometer coordinate



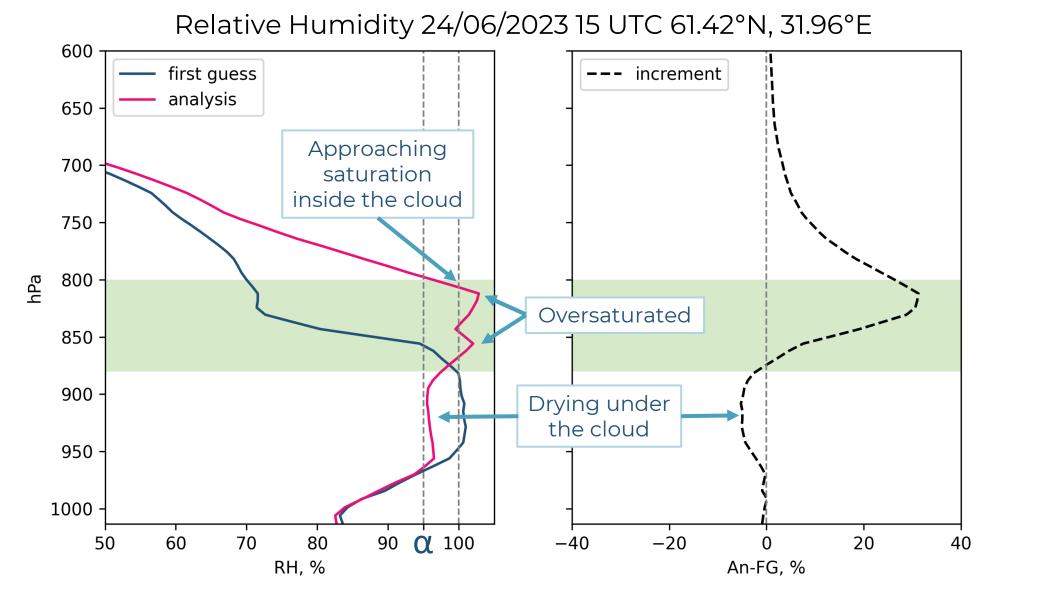




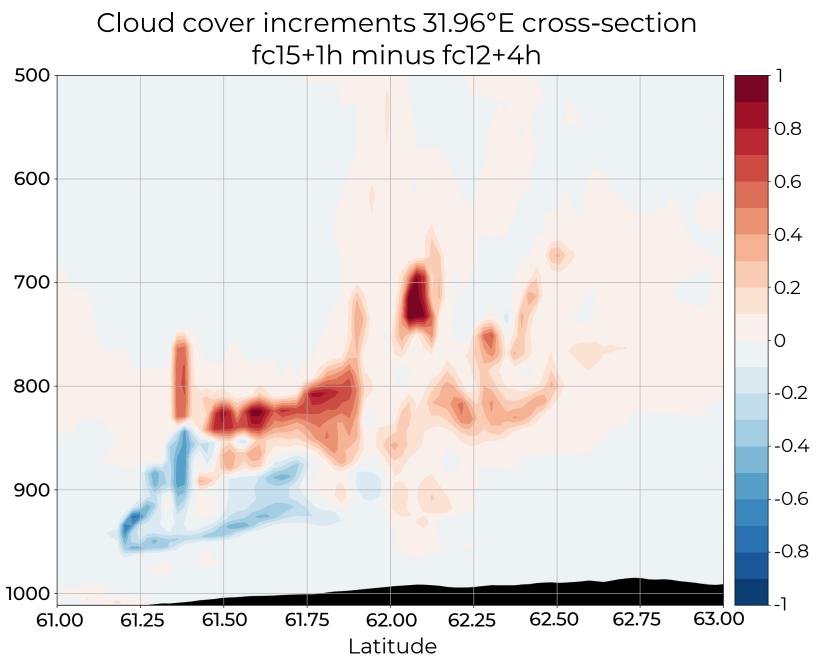
Case3 (21 UTC) - Cloud elimination First guess: cloud layer at ca. 990-970 hPa Experiment: cloud-free

Results

Case1: cloud shift



Case2: cloud creation



Low cloud cover increments fc18+1h minus fc15+4h V

model vertical levels

Penalty Function Definition

Instead of the observation operator, we construct the observation penalty function, which penalizes:

- too low RH values inside the cloud layer
- too high RH values outside the cloud layer

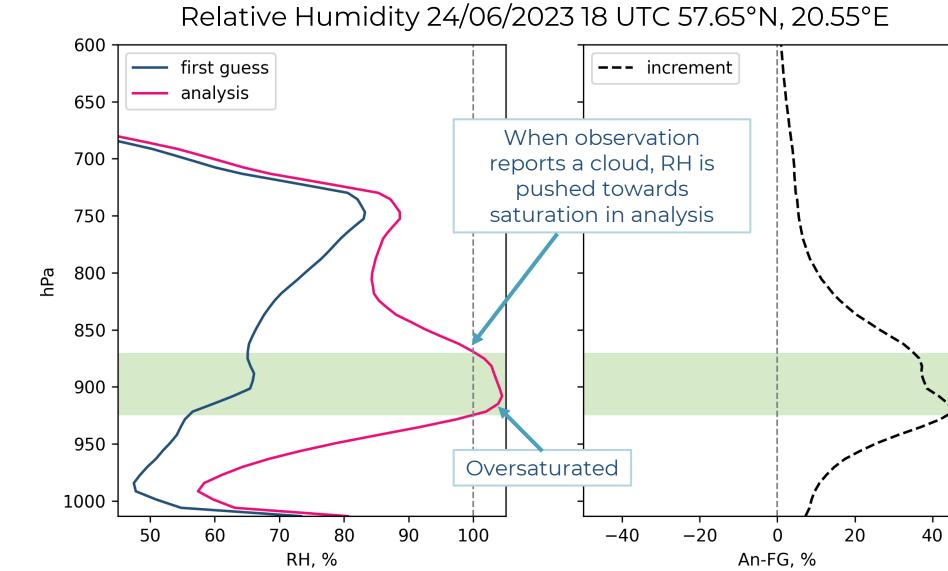
1) Constraints definition

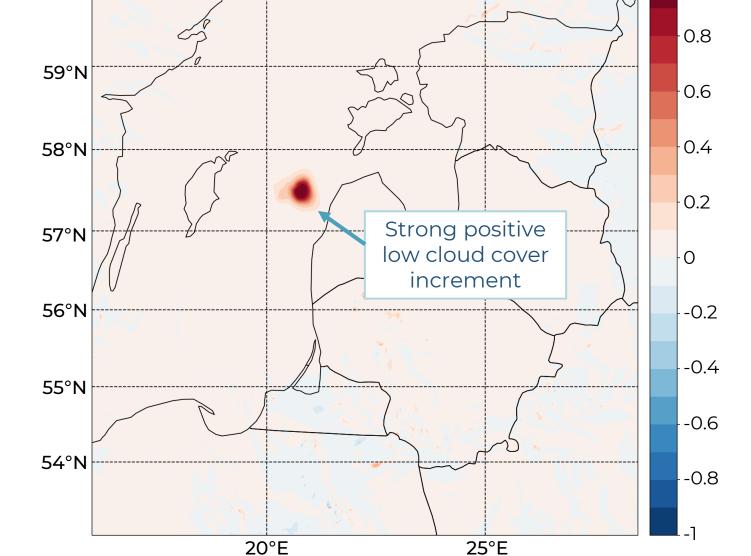
RH(vl)-1, for $vl \in [p_b, p_t]$ $P(vl) = \langle RH(vl) - \alpha,$ for $vl \notin [p_b, p_t]$ and $RH(vl) > \alpha$ for $vl \notin [p_b, p_t]$ and $RH(vl) \leq \alpha$ 0,

vl – vertical model level; where RH(vl) – relative humidity computed from the first guess p, T, q fields; α – threshold value.

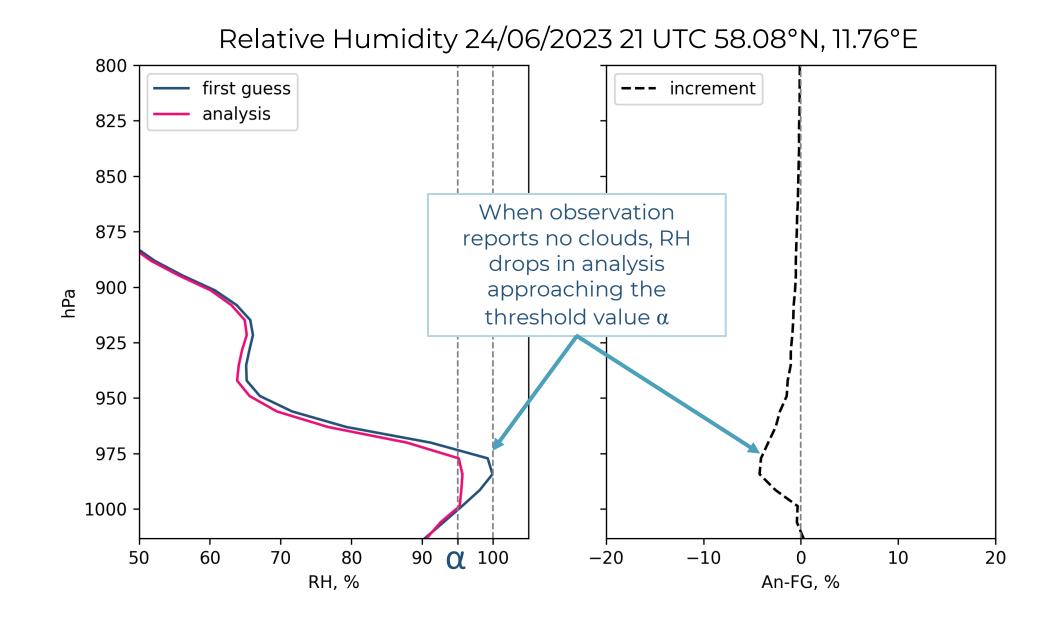
2) Observation operator \rightarrow Penalty function $H(x) = \tilde{P} = y + P$ H(x) – observation operator; where \tilde{P} – penalty function;

y – observation.

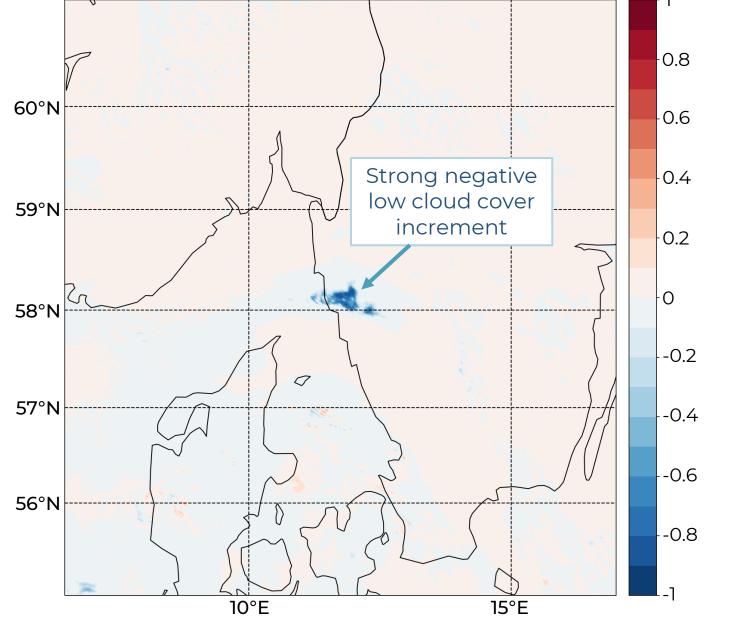




Case3: cloud elimination



Low cloud cover increments fc21+1h minus fc18+4h



3) Contribution to J_{o}

 $J_o = \frac{1}{2} [y - H(x)]^T R^{-1} [y - H(x)]$ = $\frac{1}{2} (y - y - P)^T R^{-1} (y - y - P) =$ $\frac{1}{2} (-P)^T R^{-1} (-P)$

where J_o – observation term of the cost function of the analysis; R – observation error covariances matrix.

Acknowledgements

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Conclusions

- A new scheme of cloud data assimilation is being developed based on the concept of defining the observation operator through the penalty function.
- Three single pseudo-observation experiments have been conducted using a Harmonie-Arome model setup with the cy43h2.2.1 in the HMEST25 domain.
- All three cases have demonstrated the sensitivity of RH values to the artificial cloud observations while small oversaturation is generated in the middle of the cloud layer.
- RH updates in the analyses have the according impact on the forecasted cloud cover when compared to the forecast from the previous analysis step.
- The next step for the scheme requires formulation of a cloud top height estimate in addition to the cloud base measurements from the ceilometer with subsequent real case experiments.