

# Framework for assimilating all-sky MHS radiance data in the KIM forecast system

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# Background

*“All-sky radiance assimilation improves the analysis and shorter-range forecasting of otherwise poorly observed weather phenomena as diverse as tropical cyclones and wintertime low cloud” (Geer et al., 2018).*

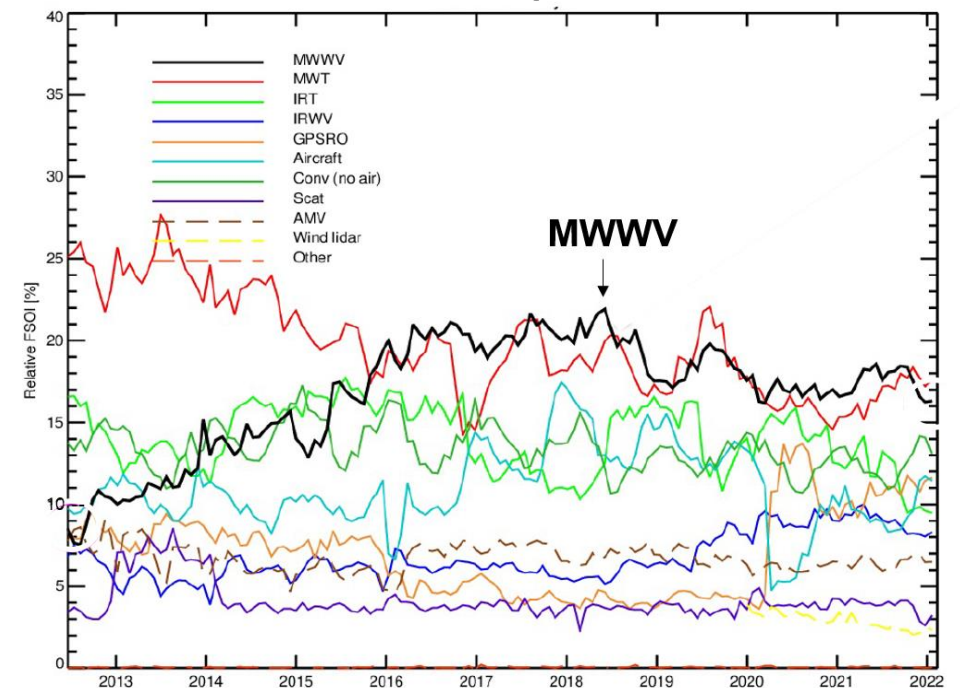
Some NWP centers already assimilate cloud- and precipitation-affected radiances operationally.

**In this talk, the framework for assimilating all-sky MW radiance data in the KIM system will be introduced.**

## ➤ Requirements for all-sky radiance assimilation

- observations
- forecast model that represents cloud and precipitation
- data assimilation system that can handle nonlinearity
- fast observation operator that represents cloud and precipitation
- observation error model

**FSOI of major observing systems in ECMWF operations**



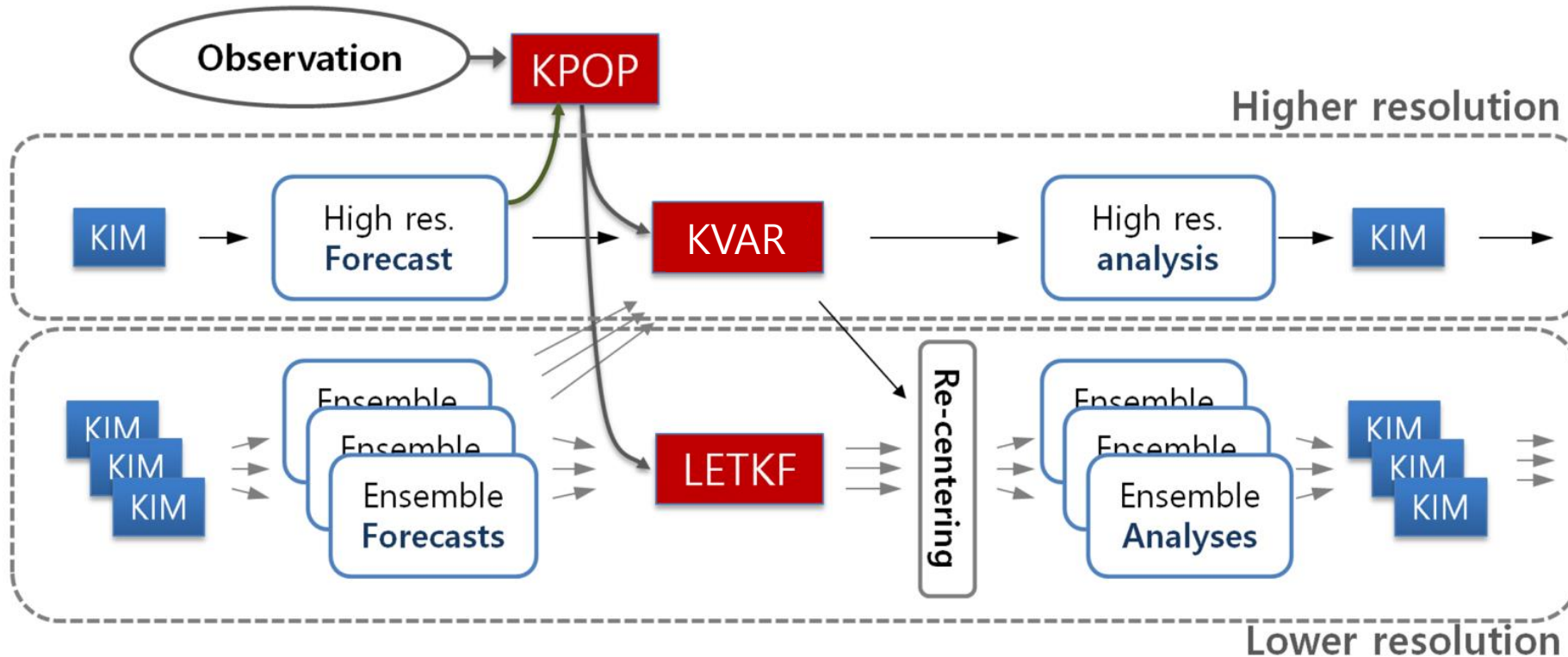
B. Ingleby (2021) ECMWF DA Training course

# Framework for all-sky DA in the KIM system

# Operational KIM system

- Ensemble DA: **50-member LETKF**

- Deterministic DA: **Hybrid-4DEnVar**



**Deterministic** resolution: NE360NP3 ~12 km  
**Ensemble** resolution: NE144NP3 ~32 km  
**Analysis** resolution: NE144NP3 ~32 km

**Ensemble B fraction:**  
 0.7 at equator, 0.45 at poles

# Observation operator

➤ RTTOV-SCATT v.13.0 is employed for the observation operator of MW satellite radiance data.

## RTTOV-SCATT v13 hydrometeors

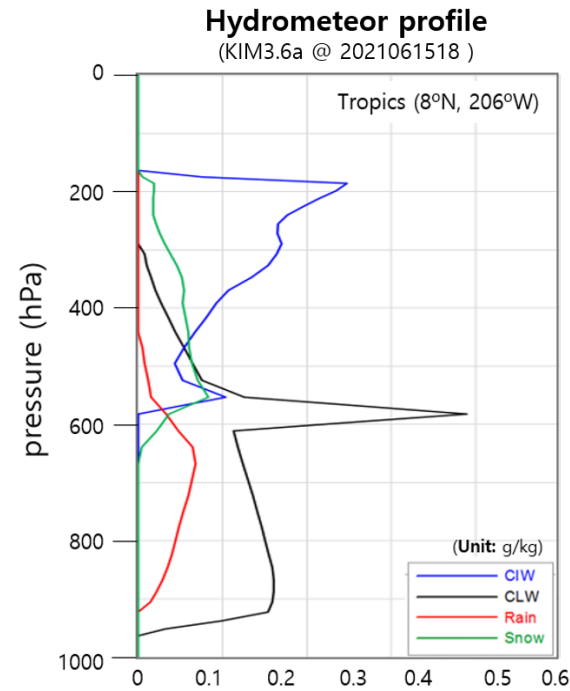
- hydro\_frac(1)** : nlevels of cloud cover (0-1)
- hydro(1)** : nlevels of rain (kg/kg)
- hydro(2)** : nlevels of frozen precipitation (kg/kg)
- hydro(3)** : nlevels of graupel (kg/kg)
- hydro(4)** : nlevels of liquid water (kg/kg)
- hydro(5)** : nlevels of ice water (kg/kg)

↕ match up

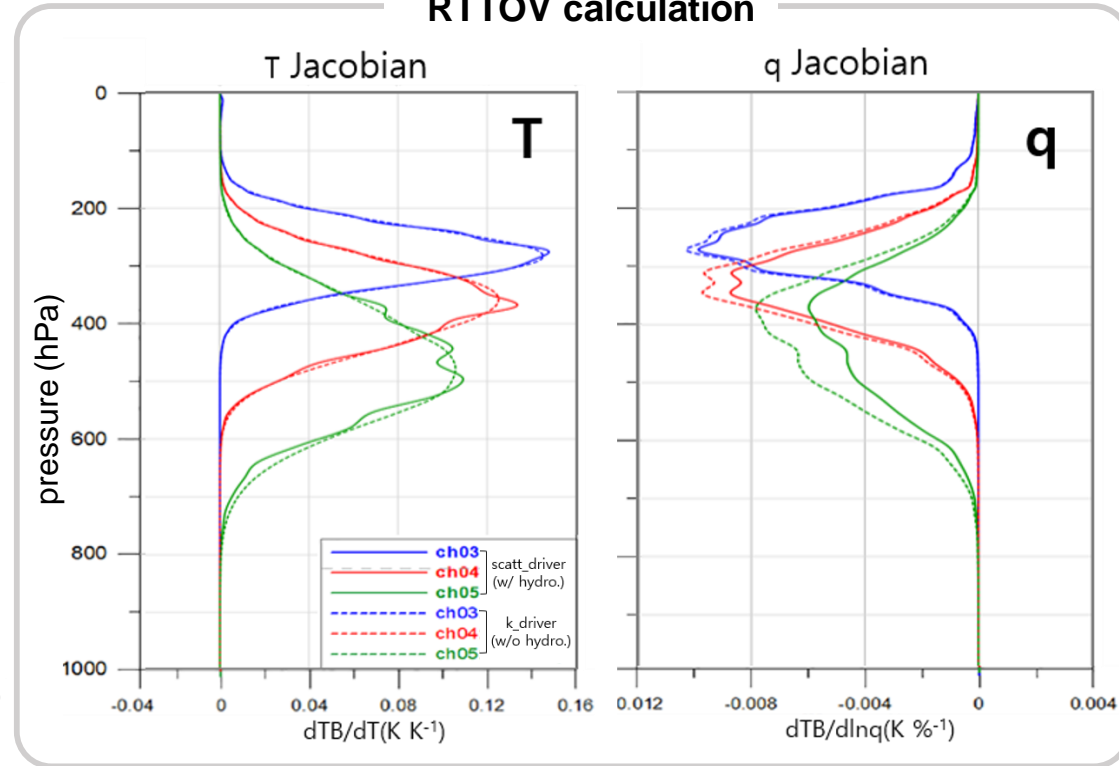
## KIM hydrometeors

- cld** : cloud fraction in atmosphere layer (0-1)
- tqr** : rain water content from mps\*, cps\*\* and scv\*\*\* (kg/kg)
- tqs** : snow content from mps, cps and scv (kg/kg)
- qg** : graupel content from mps (kg/kg) → 0.0
- tqc** : cloud liquid water from mps, cps and scv (kg/kg)
- tqi** : cloud ice content from mps, cps and scv (kg/kg)

\*mps : microphysical scheme (grid-scale, prognostic)  
 \*\*cps : convective parameterization scheme (subgrid-scale, diagnostic)  
 \*\*\*scv : shallow-convection parameterization scheme (subgrid-scale, diagnostic)



## RTTOV calculation



## Simulated TBs calculated from RTTOV

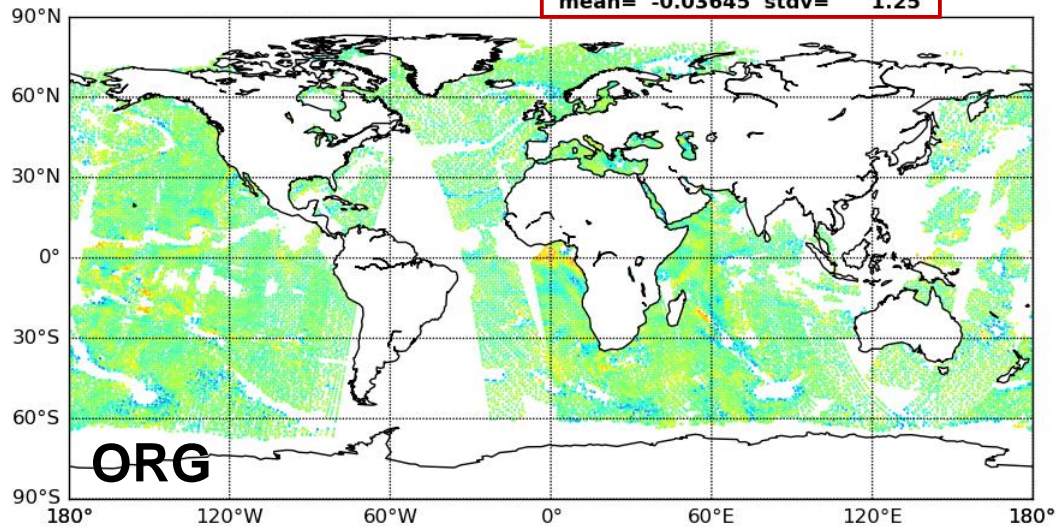
MHS	scatt_driver	k_driver
CH3	236.5	237.0
CH4	250.2	251.7
CH5	259.7	263.2

# Global distribution of O-B

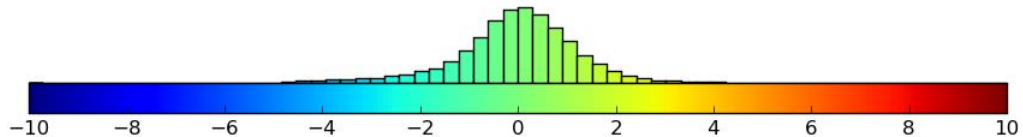
## Clear-sky assimilation

KPOP MHS ch05: O-B @ 2021061518

mean= -0.03645 stdv= 1.25



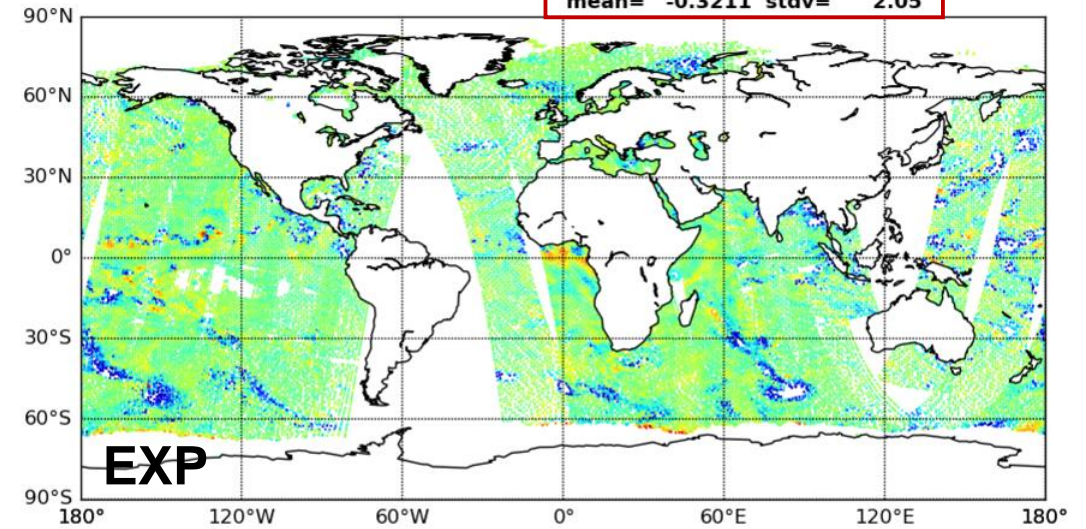
Total: 38539 (100%)



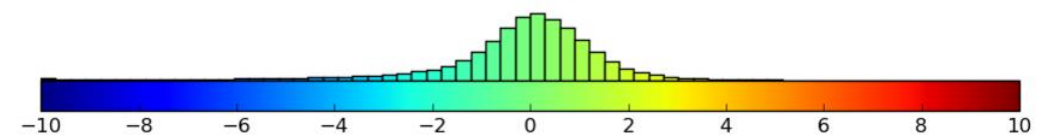
## All-sky assimilation

KPOP MHS ch05: O-B @ 2021061518

mean= -0.3211 stdv= 2.05



Total: 45726 (119%)



**All-sky  
assimilation**

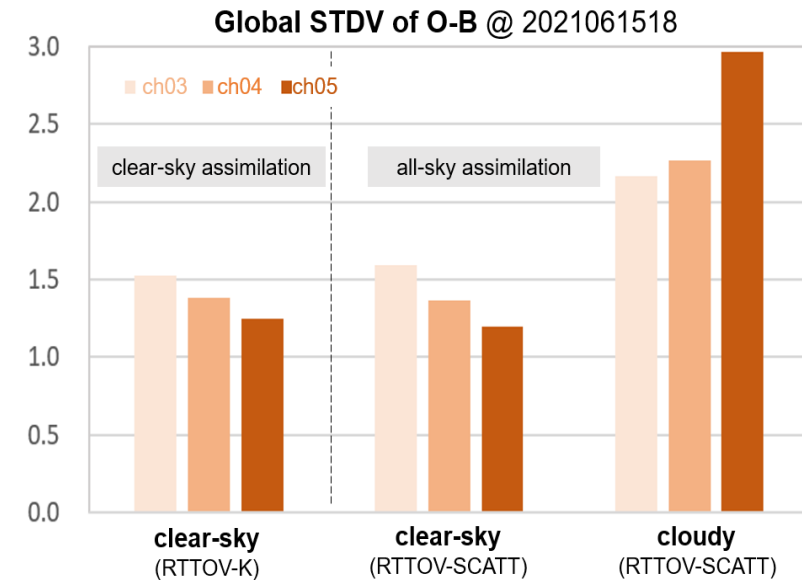
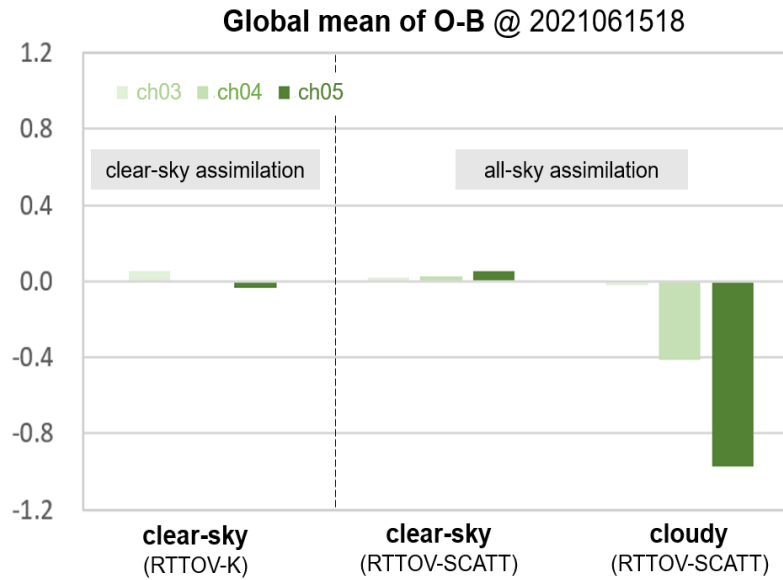
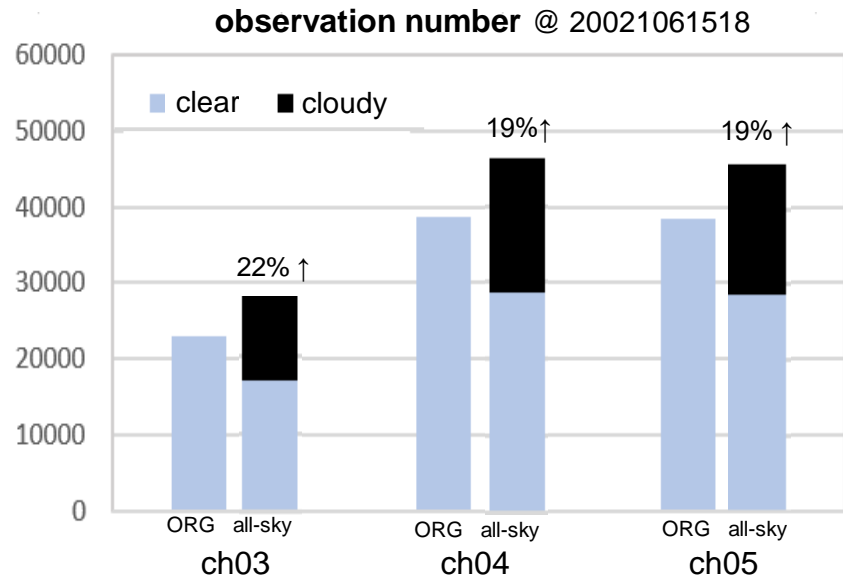
### Quality control

- removal of cloud masking
- O-B check as a function of symmetric cloud amount

### Bias correction

- selection of clear pixels to calculate BC coefficients
- clear-pixel : symmetric cloud amount  $\leq 0$

# O-B statistics



Approximately 20% more observations are assimilated in all-sky approach.

In both experiments, O-B statistics at clear-pixels are similar.

Mean and standard deviation of O-B are relatively large in cloudy pixels (especially, MHS lowest channel), which will be considered in observation error.

# Observation error

## Geer et al. (2014) method

- observation error calculation as a function of symmetric cloud amount (average of **SI\_bkg** and **SI\_obs**)

$$g(C_{SYM}) = g_{clr} + (g_{cld} - g_{clr}) \left( \frac{C_{SYM} - C_{clr}}{C_{cld} - C_{clr}} \right)^2 \quad \in C_{clr} < C_{SYM} < C_{cld}$$

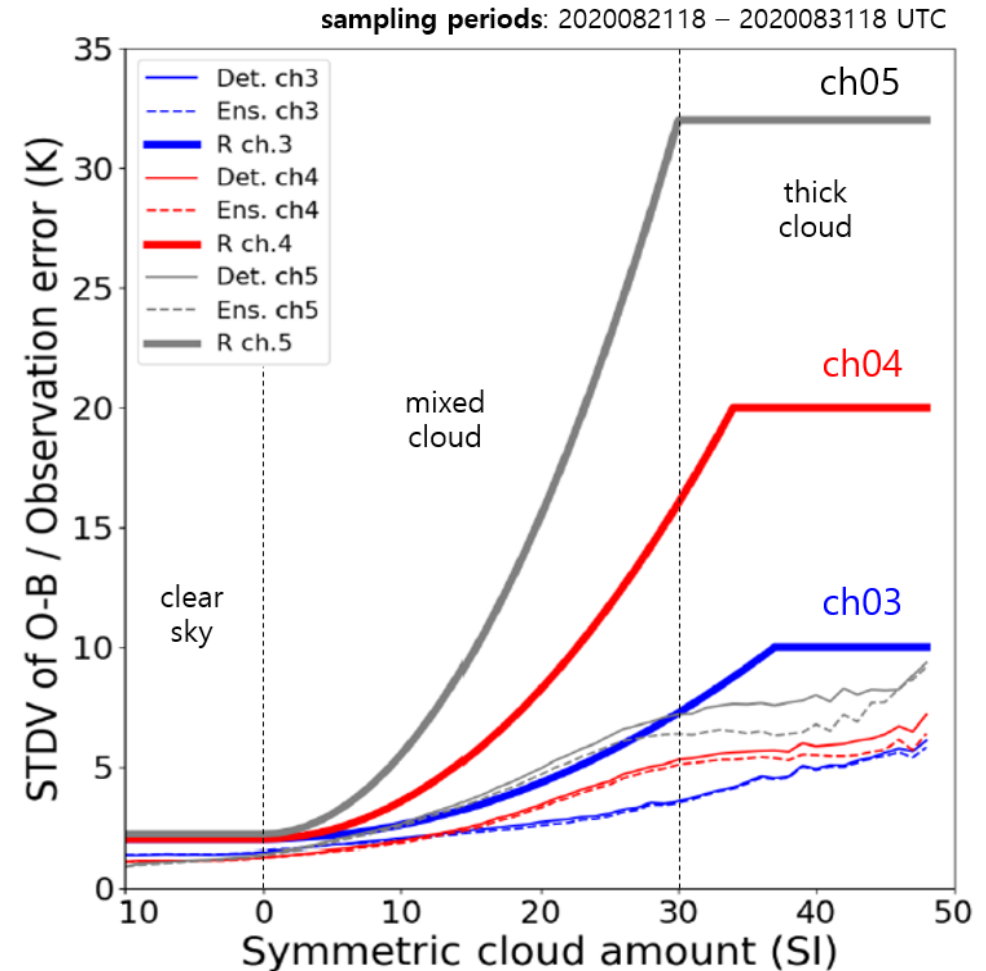
$g(C_{SYM})$  : total error     $C_{SYM}$  : symmetric cloud predictor

(e.g.) MHS ch05

- clear sky : obs\_error = 2 (K)
- mixed cloud : obs\_error = 2.2+(32-2.2)(avg\_SI/30)\*\*2 (K)
- thick cloud : obs\_error = 32 (K)

※ Scattering Index (SI) is used for cloud predictor.

※ For beneficial impact of all-sky assimilation, optimization of observation error will be needed.





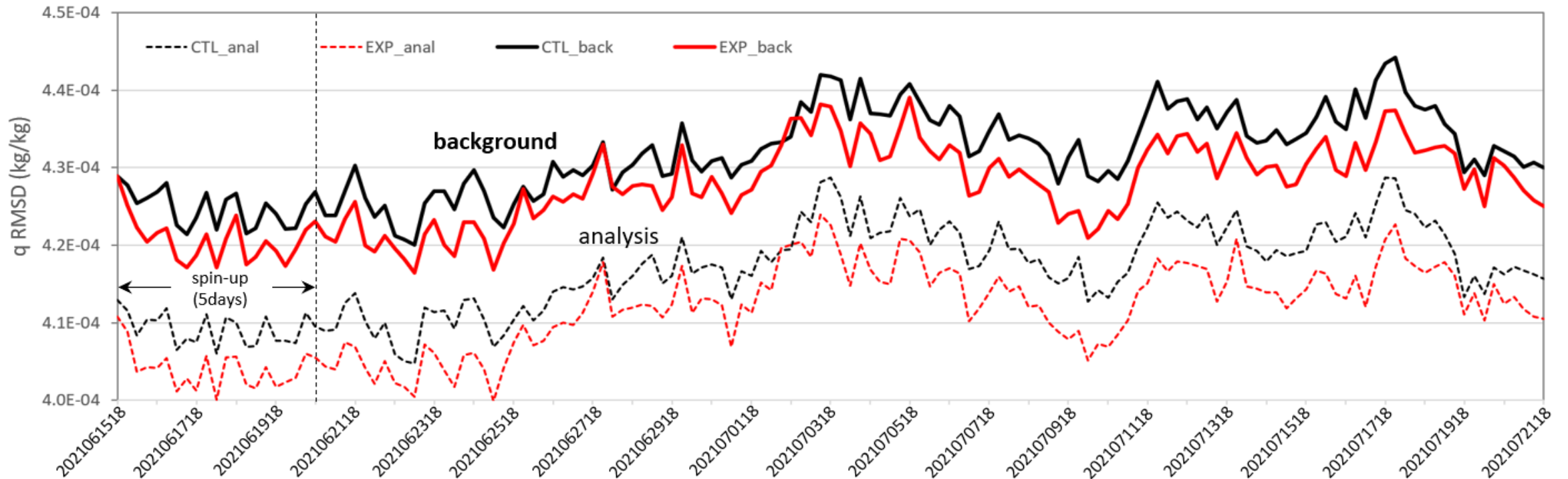
# Preliminary results

# Cycled analysis and forecast experiment

- **Period** : 2021061518 - 2021072118 UTC (36 days)
- **Used observation** : Sonde, Surface, Aircraft, GPS-RO, Scatwind, AMV, AMSU-A, MHS  
( **CTL** : clear-sky MHS, **EXP** : all-sky MHS )
- **Verification** : IFS analysis ( $0.25^\circ \times 0.25^\circ$ )

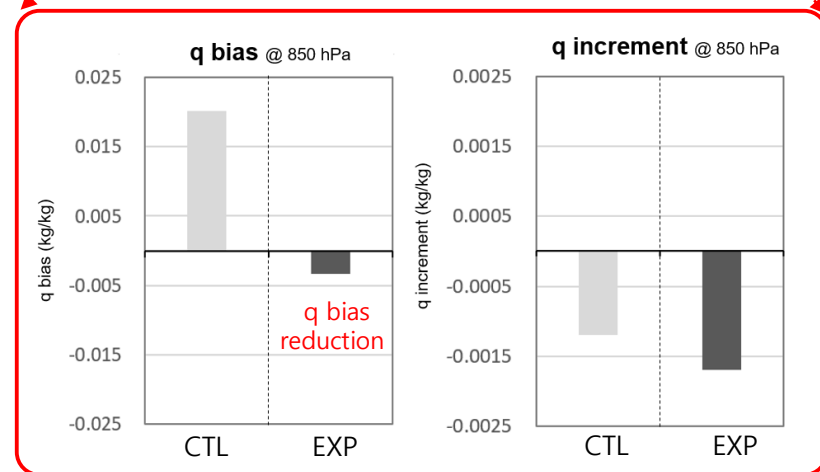
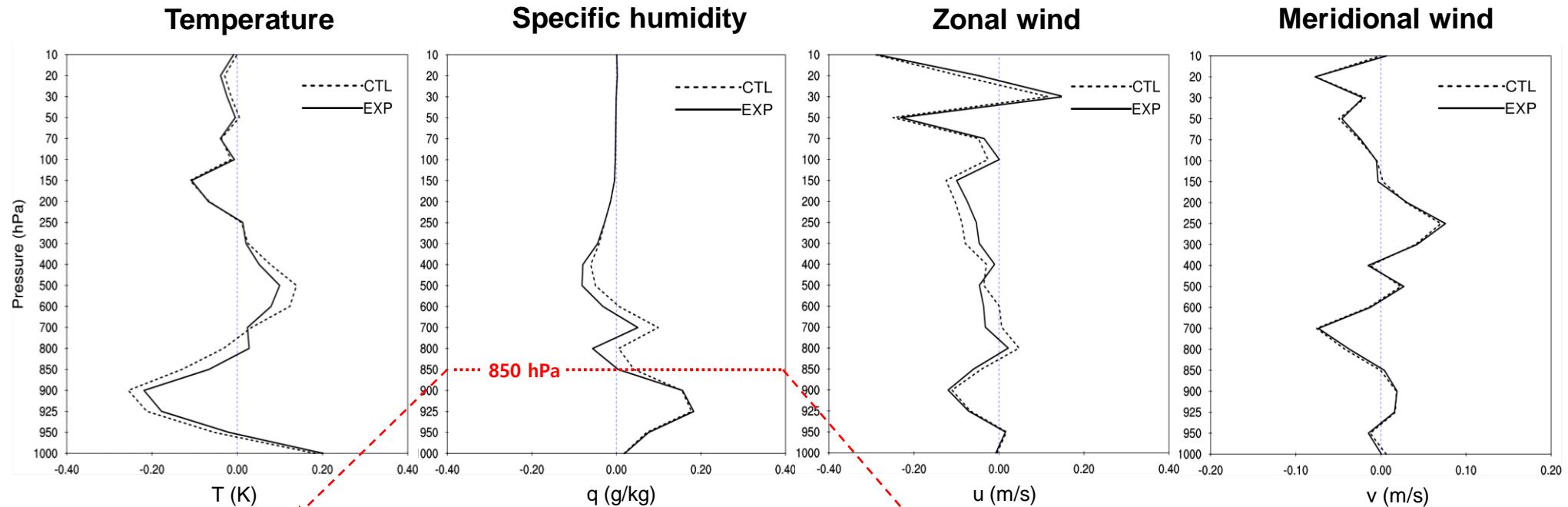
KIM\_v3.7 (NE180NP3 ~ 25 km)  
KVAR\_r224 (NE144NP3 ~ 32 km)  
KPOP\_r564 (NE180NP3 ~ 25 km)

Timeseries of global **specific humidity** RMSD



# Vertical profile of background bias against IFS analysis

\* Period: 2021062100 – 2021072118 UTC (31 days)



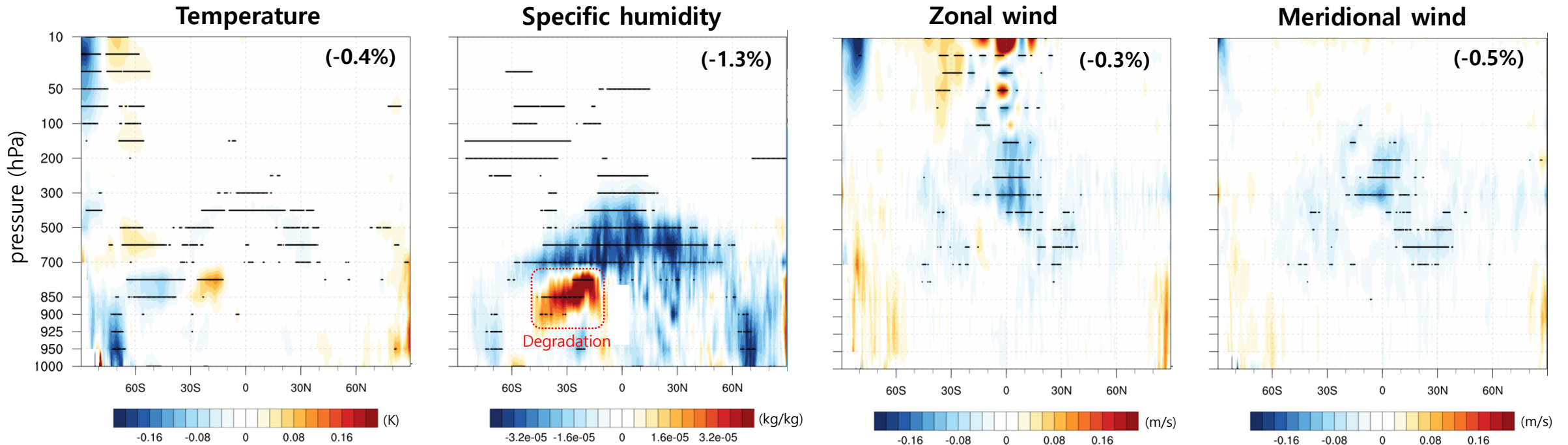
In the cycled experiments, all-sky assimilation of MHS (EXP) gives a beneficial impact on mass and wind background fields.

The background of EXP has become drier in the mid troposphere.

# Zonal plot of analysis RMSD against IFS analysis

\* Period: 2021062100 – 2021072118 UTC (31 days)

Difference of analysis RMSD (EXP – CTL)



\* Blue color means beneficial analysis impact of all-sky assimilation.

\* Small black dots show a 95% significant difference, verified by a *t*-test, between CTL and EXP.

Overall, assimilation of MHS observations in cloudy regions gives a beneficial impact on T, q and wind analyses.

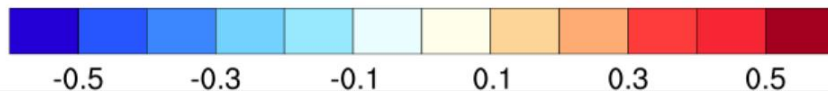
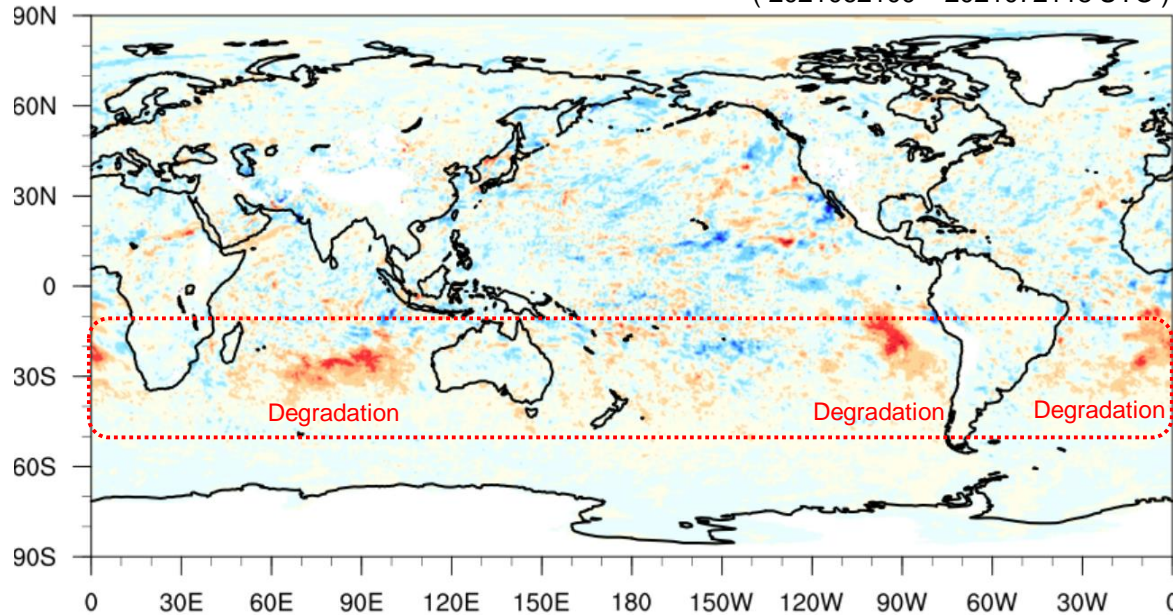
However, the degradation of q analysis in the SH midlatitudes (10-50°S) for the lower tropospheric levels is shown.

# Mean field of specific humidity at 850 hPa

\* Verification : IFS analysis

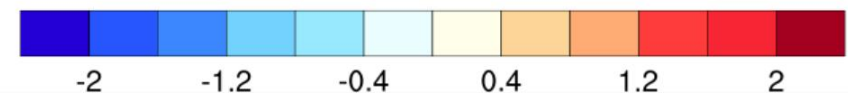
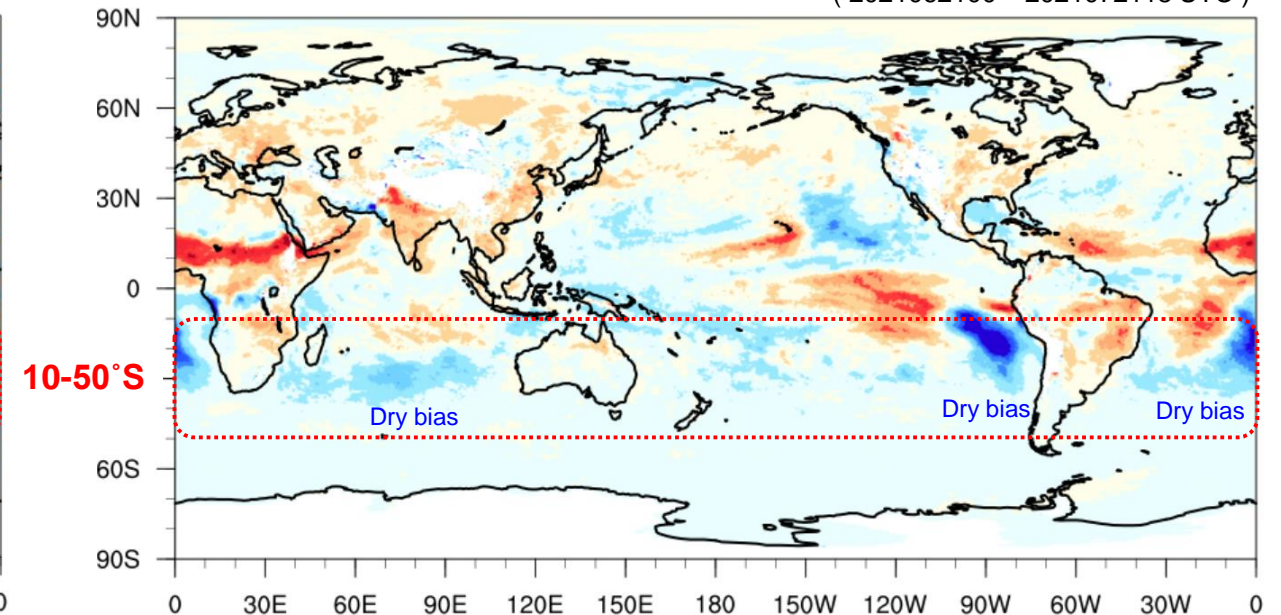
### Difference of q analysis RMSD (EXP-CTL) @ 850 hPa

( 2021062100 – 2021072118 UTC )



### EXP : q background bias @ 850 hPa

( 2021062100 – 2021072118 UTC )

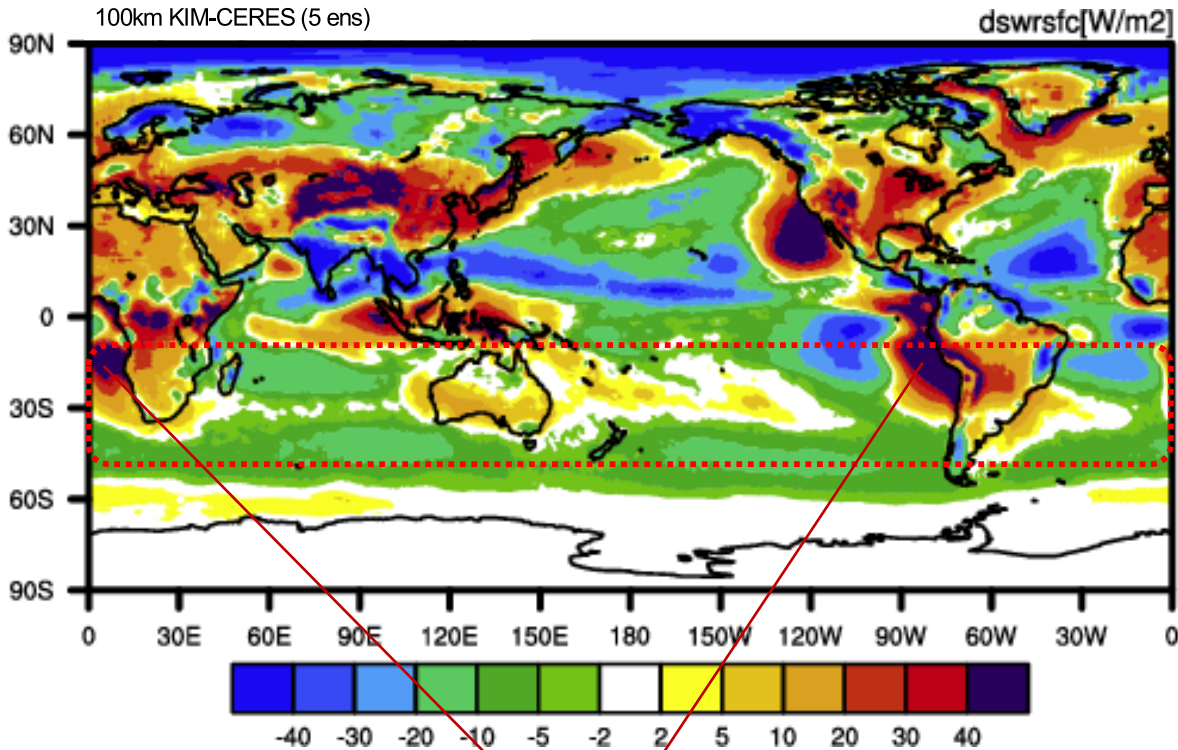


The degradation of q analysis from all-sky DA (EXP) appears in dry-biased background areas.

# Cloud estimation

## Bias of downward SW radiation

JJA mean 2011-2020  
100km KIM-CERES (5 ens)



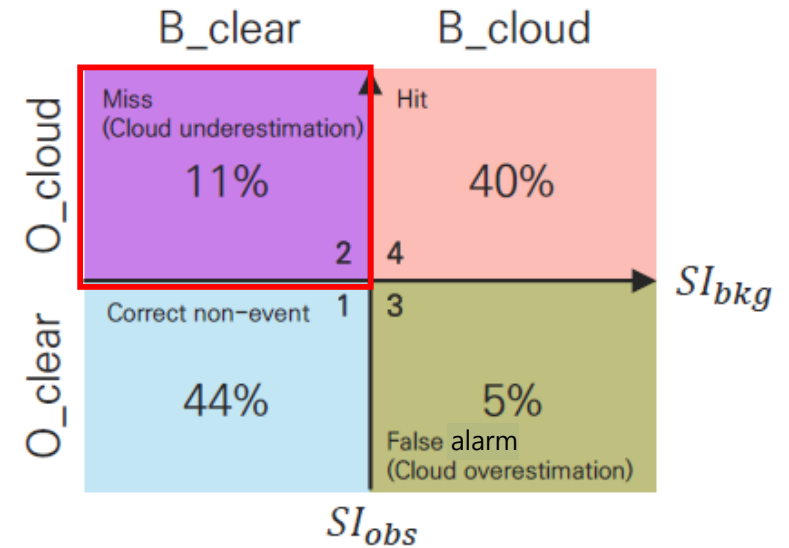
cloud underestimation

Thanks to E-H Lee

In the SH midlatitudes, the areas with dry bias correspond to areas with underestimated cloud.

## Cloud mismatch

@ MHS ch05



\* Period: 2021062500 – 2021073118 UTC

Thanks to H-B Jeong

In the cloud underestimated region (O<sub>cloud</sub>/B<sub>clear</sub>), all-sky pixels should be assimilated more conservatively. (e.g., observation with inflated observation error)

→ see H-B Jeong's Poster (9p.03)

# Summary

- The Korean Integrated Model (KIM) forecast system has been extended to assimilate all-sky radiances from microwave satellite sensors (e.g., MHS and ATMS).
- RTTOV-SCATT version 13.0 (observation operator) was implemented to assimilate the MW water vapor channels over the ocean.
- The observation error model of Geer et al. (2014) has been adopted for all-sky assimilation in the hybrid-4DEnVar (KVAR) system.
- Overall, assimilation of MHS observations in cloudy regions gives a beneficial impact on T, q, and wind analyses in the cycled analysis and 6-hr forecast experiments.
- However, it is necessary to improve the analyses in the SH midlatitudes, because specific humidity analyses are degraded in dry-bias areas.

# THANK YOU

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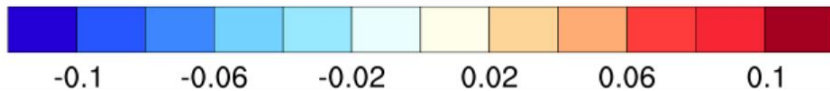
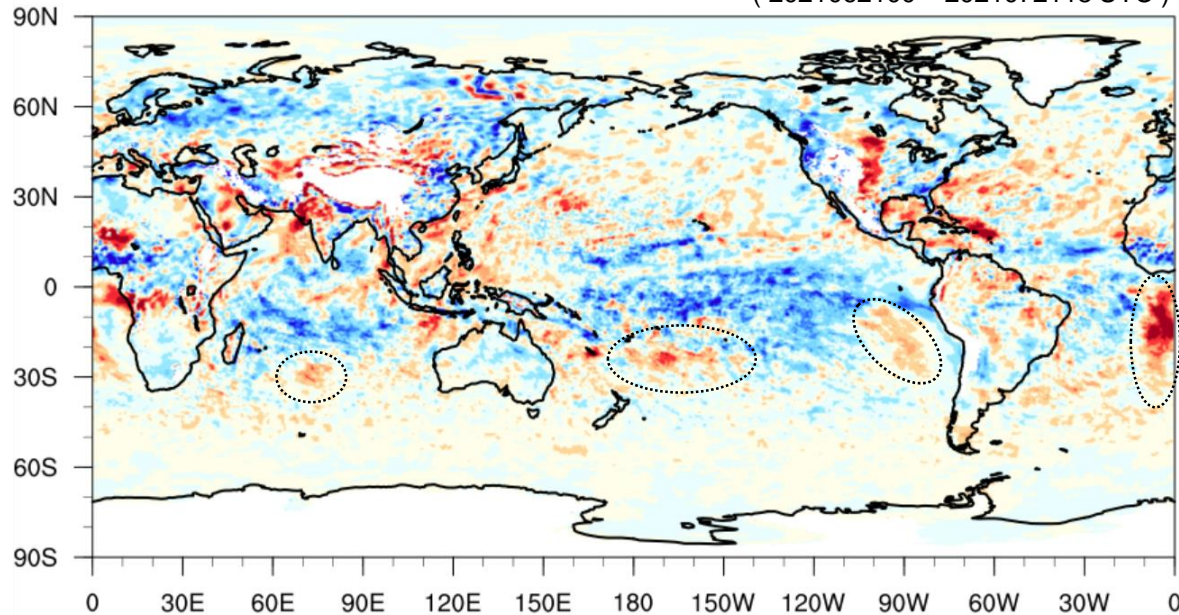


# q analysis increment at 850 hPa

Back-up

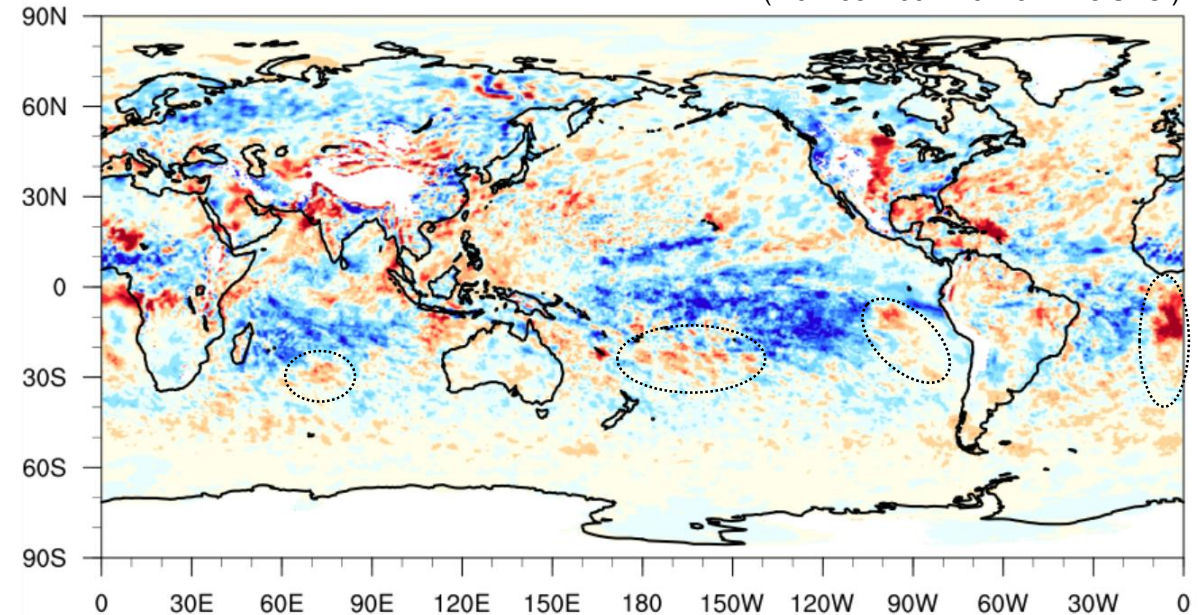
CTL : q analysis increment @ 850 hPa

( 2021062100 – 2021072118 UTC )



EXP : q analysis increment @ 850 hPa

( 2021062100 – 2021072118 UTC )



Although the global distributions of q analysis increment for both experiments are similar, the magnitude of positive (negative) q increment is relatively weaker (stronger) in the EXP.