



Applications of Microwave Satellite Data to KMA LDAPS

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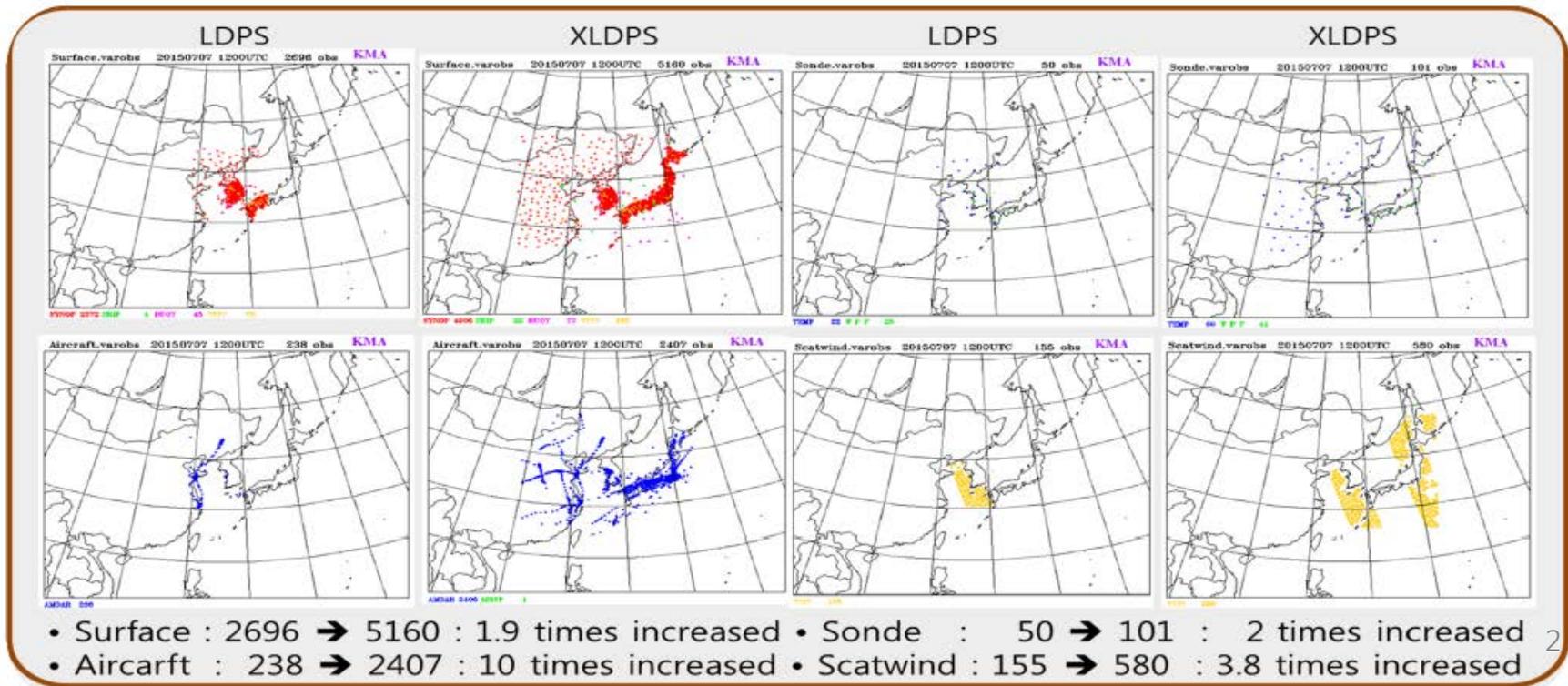
Objectives

- ❖ Introducing satellite data for synoptic information around sparse data area
- ❖ Assimilation of moisture information to the convective-scale model

Observation current usage in LDAPS

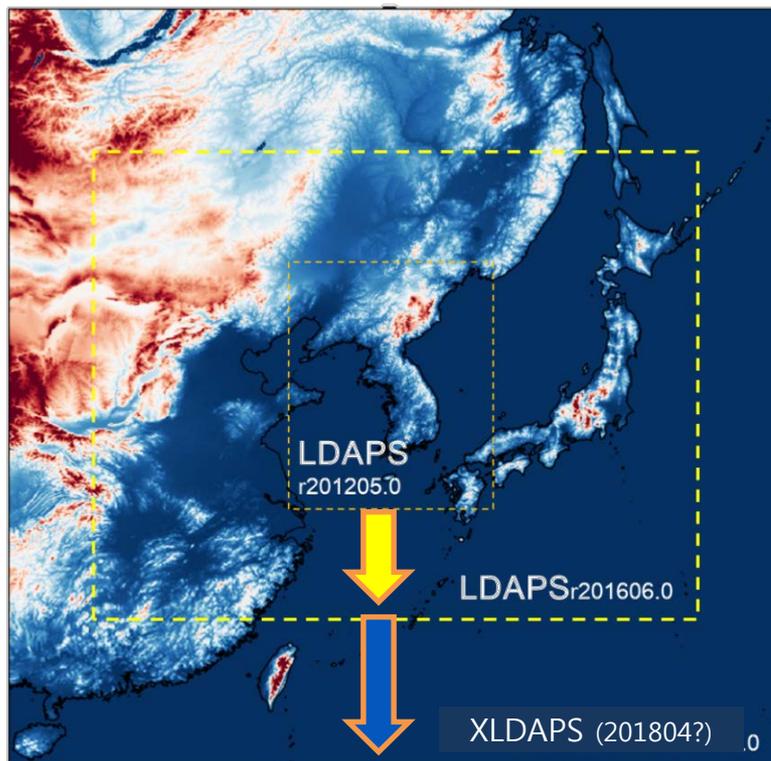
- **SONDE**(temp, pilot, windprofiler), **SURFACE**(synop, ship, buoy, metar), **AIRCRAFT**(amdar), **RADAR**(radial velocity), **SCATWIND**(ASCAT)

Lack of available observation, needs of satellite DA



KMA Convective-scale Model

❖ LDAPS (Local Data Assimilation and Prediction System)

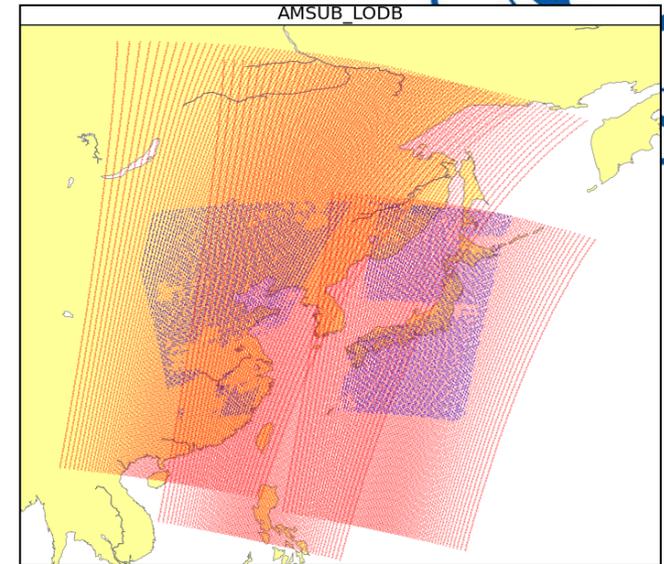


- **Model**
 - UM **vn10.1k** (ENDGame)
- **Area, resolution**
 - grid number : **1,598** (E-W) X **1,718** (S-N)
 - resolution : 1.5 ~4 km (Variable grid), 70 levels
DA 3 km
- **Forecast length (cycle)**
 - **36 hours** (3 hourly)
- **DA system: 3DVAR(FGAT) with IAU**
 - surface, sonde, wind profiler, aircraft, radar, scatwind (current) \pm 90 min cutoff time
 - **AMSU-B, ground GNSS, TC Bogus** will be added next operation
- **In Operation** since July 2017

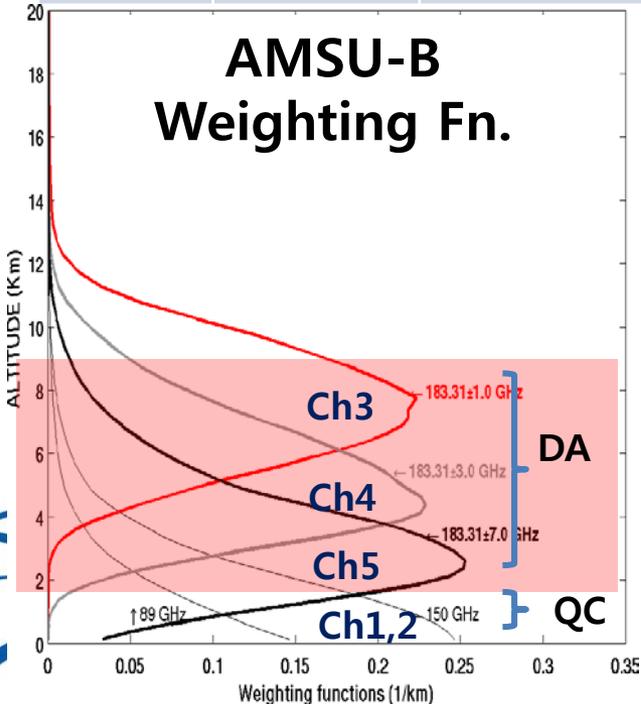
AMSU-B Data

ATOVS(MHS) on board NOAA-18, 19, METOP-A, B

Sensor		FOV [km] (at nadir)	Distance [km]	
ATOVS	HIRS	10.0	42.2	IR Sounder
	AMSU-A	47.6	52.7	MW Temp Sounder
	AMSU-B (MHS)	15.9	17.6	MW Humidity Sounder



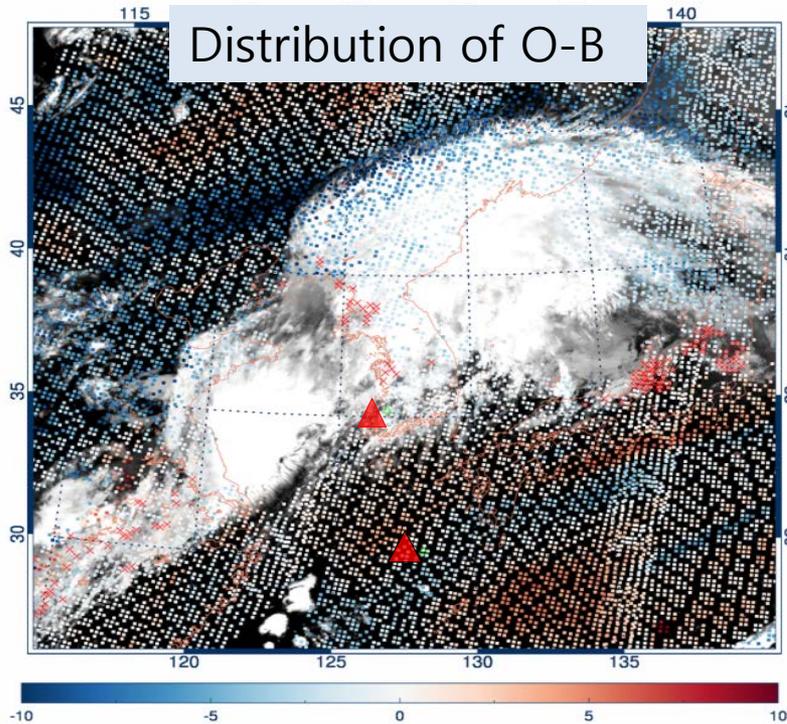
AMSU-B Weighting Fn.



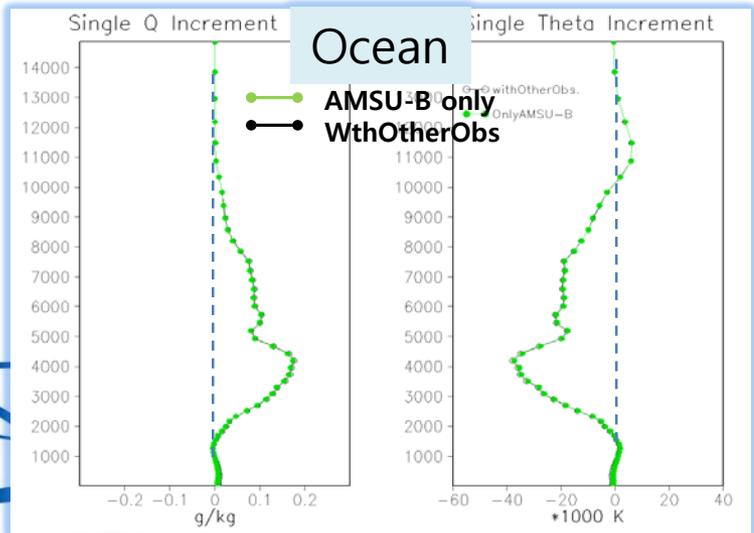
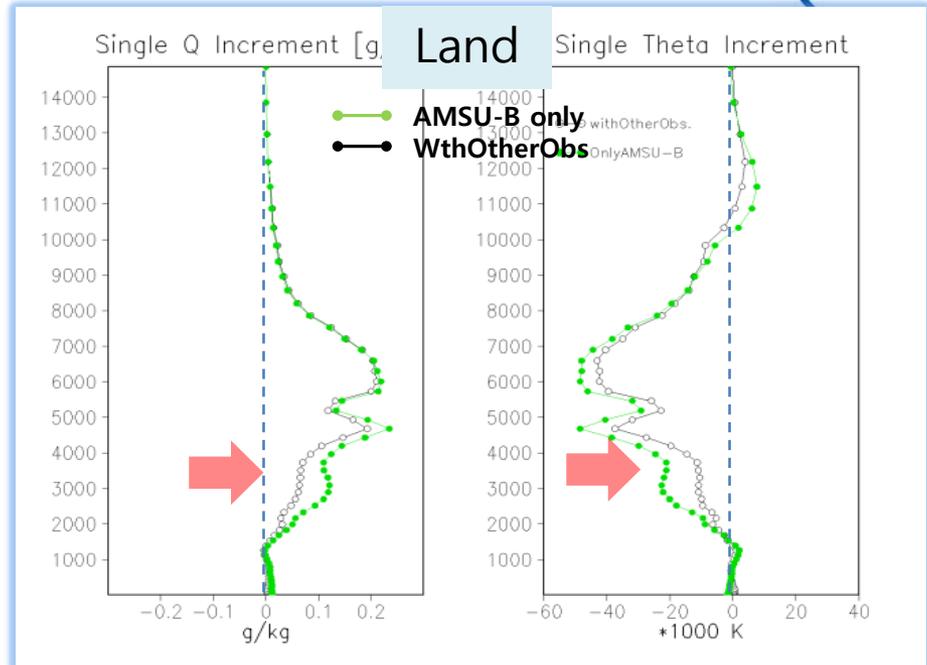
- ✓ Full-resolution AMSU-B received directly from NMSC of Korea
- ✓ Channel 1, 2 – semi-window channels, affected from surface and ice phased cloud (using for QC)
- ✓ Channel 3, 4, 5 – near 183GHz H₂O absorption line (Profiling the WV in troposphere)

AMSU-B Single Observation Test

Distribution of O-B



Analysis Increments Profiles of single AMSU-B with or without conventional data

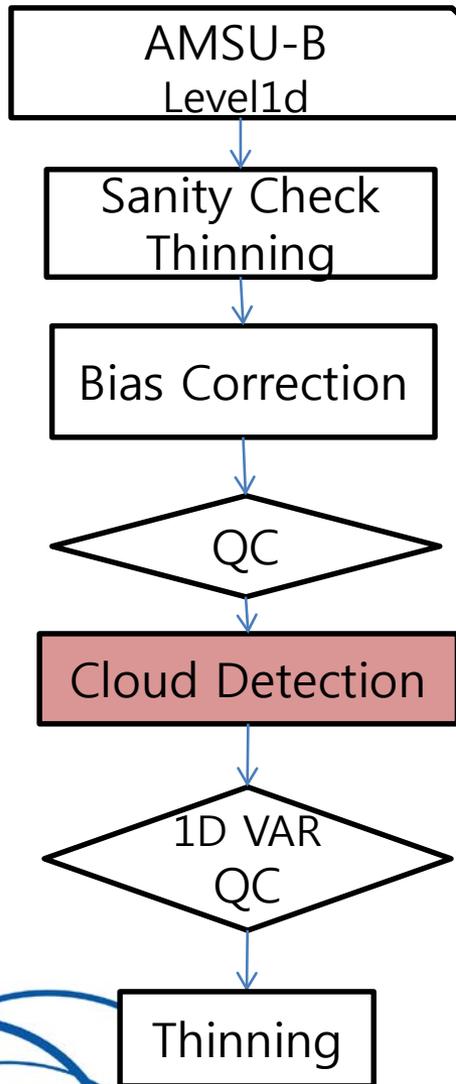


- ✓ $O-B < 0 \rightarrow$ Humidity(+), Temp.(-)
- ✓ The impact of AMSU-B is decreased but charged over 90 % of total increments over land.
- ✓ Over ocean and upper troposphere, the influence of AMSU-B is dominant.

→ We need accurate preprocess to get uncontaminated data.

OPS - Cloud Detection

Preprocess in OPS



Cloud Detection

- QC algorithm for detecting cloud-affected radiances is limited by MHS data itself

1. Detect Cirrus : reject CH4, 5

$$\text{Cloud Cost Fn. } J_c = \frac{1}{2} \Delta T \cdot \{HBH^T + R\}^{-1} (\Delta T)^2$$

(ΔT : O-B ch3,4,5 , H : TL Operator,
B: BgErr Cov., R: ObsErr Cov.)

2. (EXP1) Detect rain : reject all

$$\text{Ocean SI} = T_{b,1} - T_{b,2} - (a + b \times \theta) \quad (\theta : \text{latitude})$$

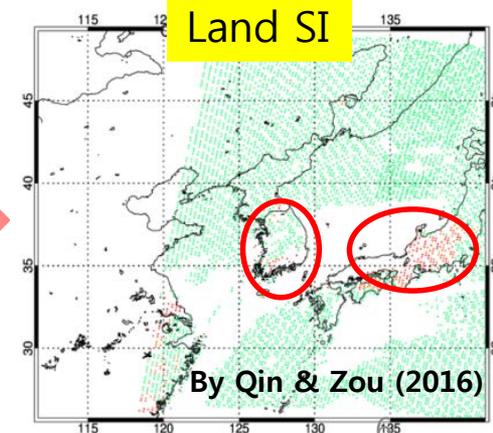
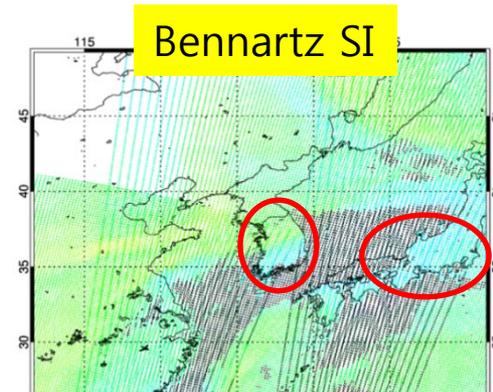
- It is not working on land

2-1. Detect Rain over Land

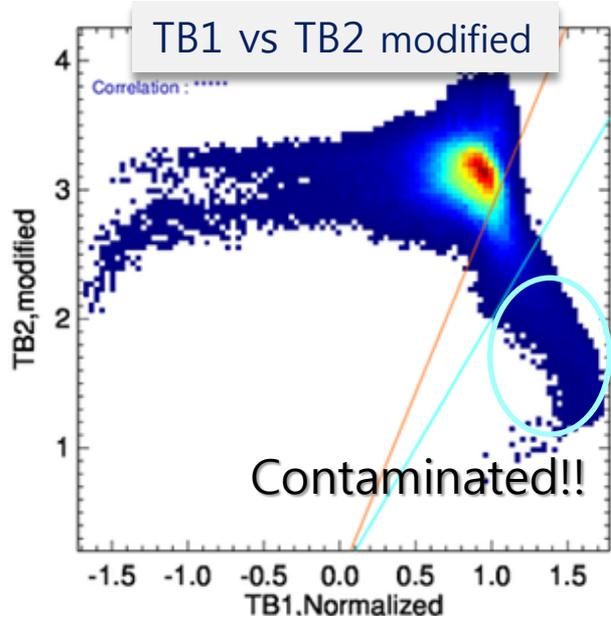
(EXP2) If cirrus detected, **reject all channels**

(EXP3) Detect rain **using Land SI***

* Qin & Zou(2016)-Development and Initial Assessment of a New Land Index for Microwave Humidity Sounder Cloud Detection



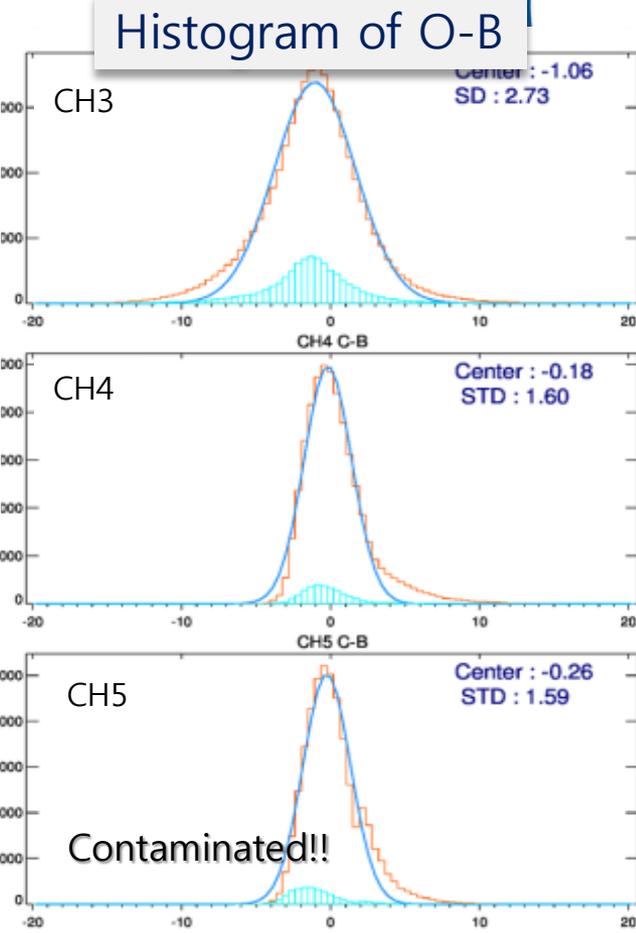
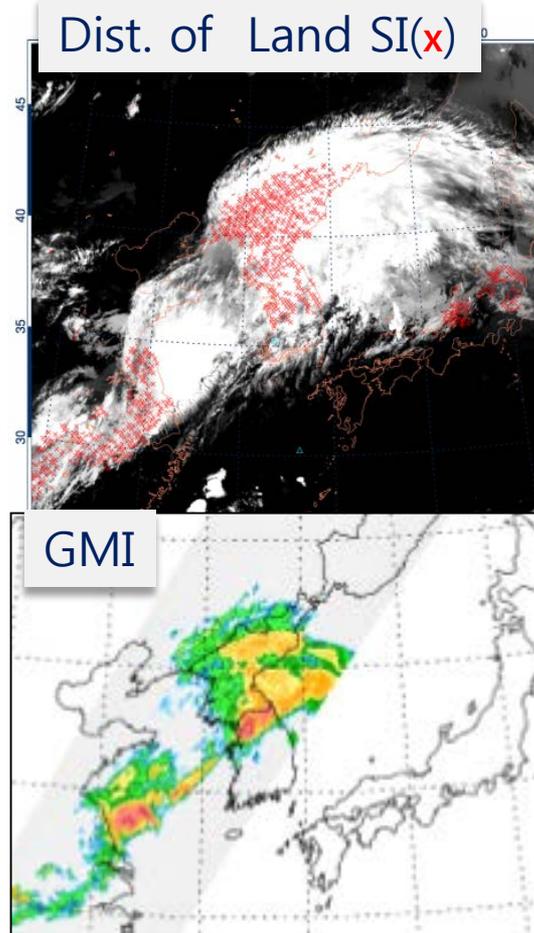
Land Scattering Index



$$L_{index} = \frac{T_{b,1}^{normalized}}{\frac{1}{2} \left(\frac{T_{b,2}^{obs}}{100} - 1 \right)^3}$$

$$(T_{b,1}^{Normalized} = \frac{T_{b,1} - \mu}{\sigma},$$

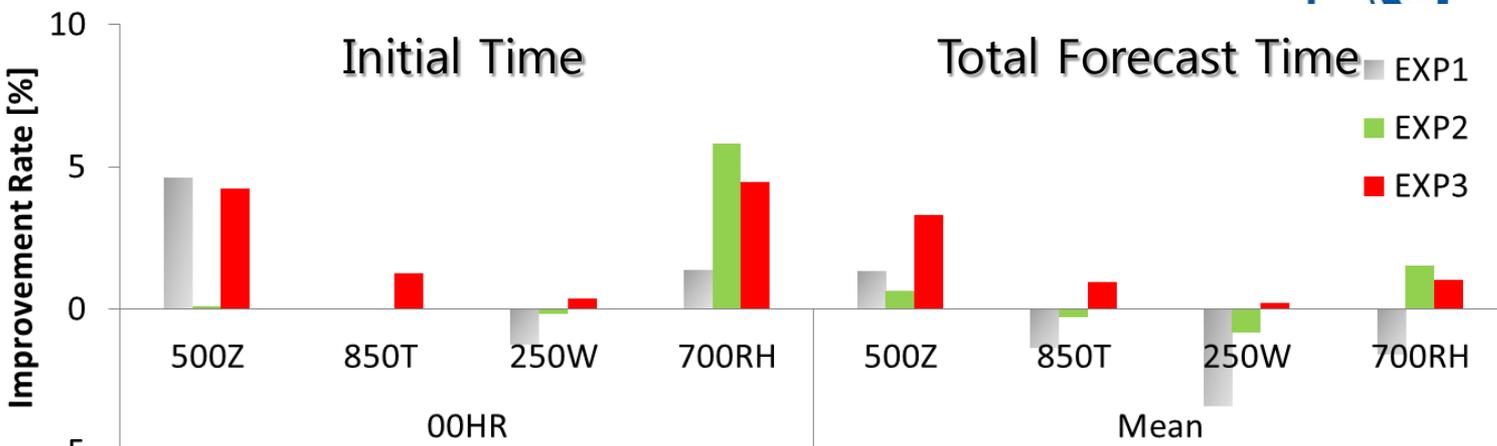
$$\mu = \frac{1}{5} \sum_{i=1}^5 T_{b,i}, \sigma = \sqrt{\frac{1}{5} \sum_{i=1}^5 (T_{b,i} - \mu)^2}$$



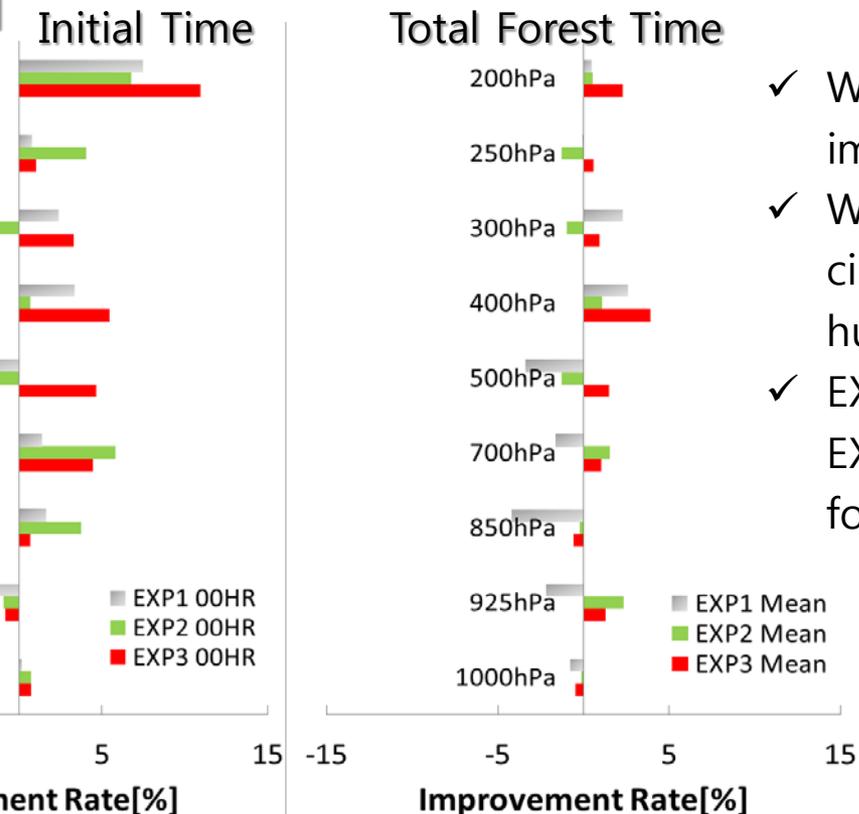
- In cloudy area, the standard deviations(σ) of MHS channel become smaller and the $T_{b,1} - \mu$ are not lower than clear sky. \rightarrow Land SI increased
- The contaminated data show negative O-B signals at lower channel.
- When remove it, the STD of departure become 1.59 to 1.55 at CH5 and the bias of departure closes to 0 at CH3 (-1.06 \rightarrow -0.28)

Model Impact Result

Verification
(with Sonde)



Humidity Field



- ✓ When adding Land SI(EXP3), it shows improvements of main fields.
- ✓ When rejecting all channel under cirrus(EXP2), it shows improvement of humidity fields at lower troposphere.
- ✓ EXP3 shows better performance than EXP2 in humidity fields with total forecast time and all height.

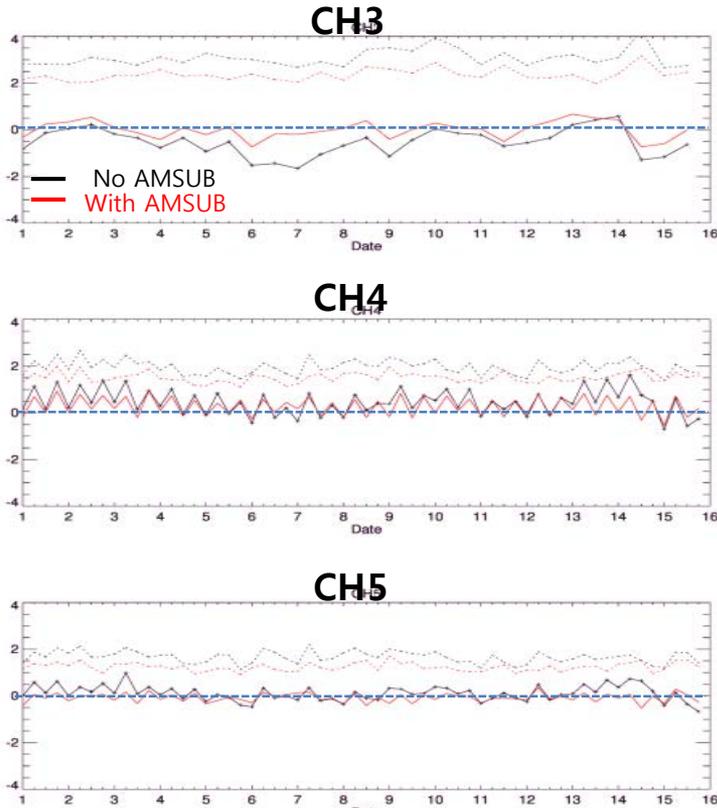
Improvement Rate [%]

$$= \left[1 - \frac{RMSE(EXP)}{RMSE(OPER)} \right] \times 100$$

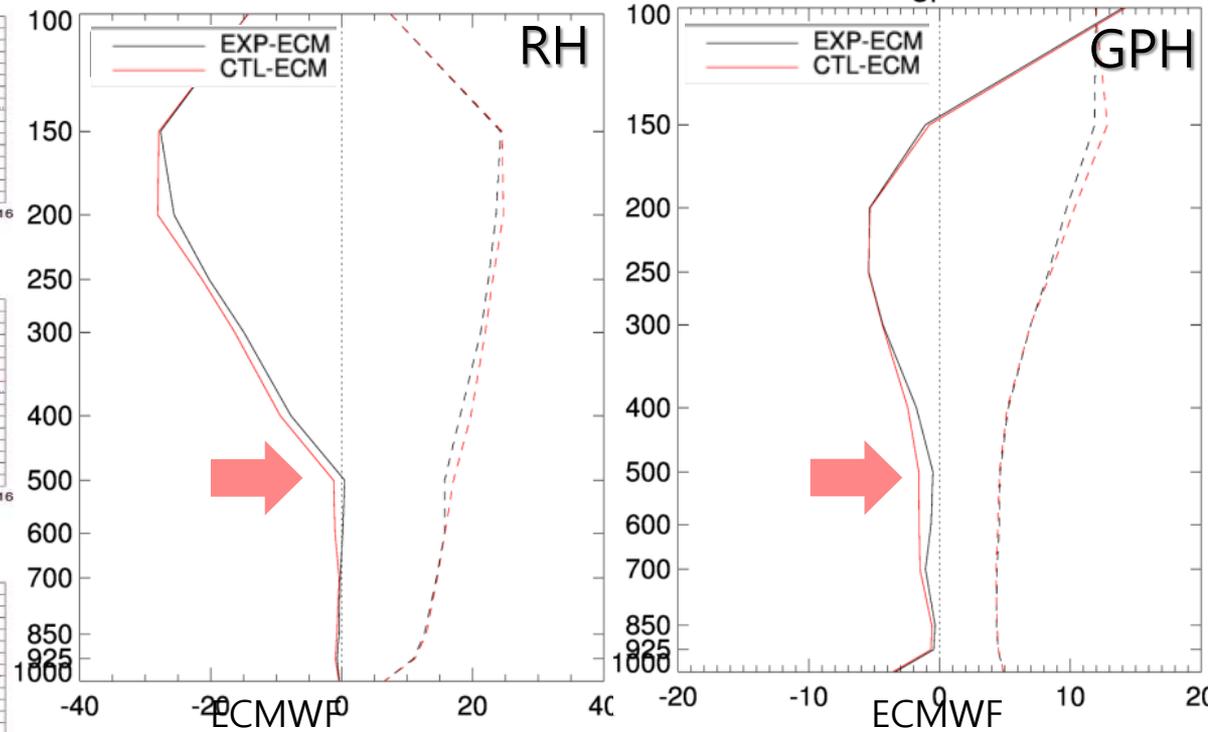
+ : improved , - : degraded

Model Impact Result

O-A (cold vs AMSU-B)



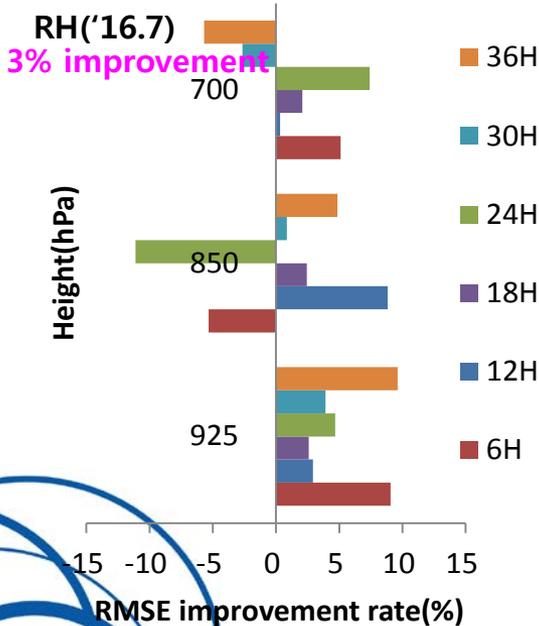
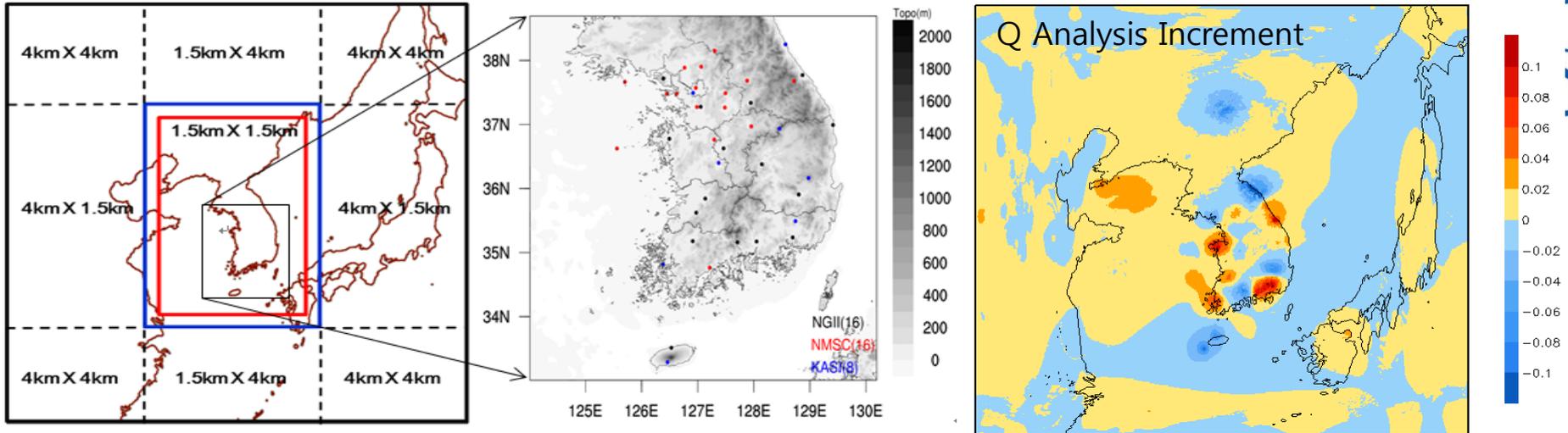
Comparison with ECMWF



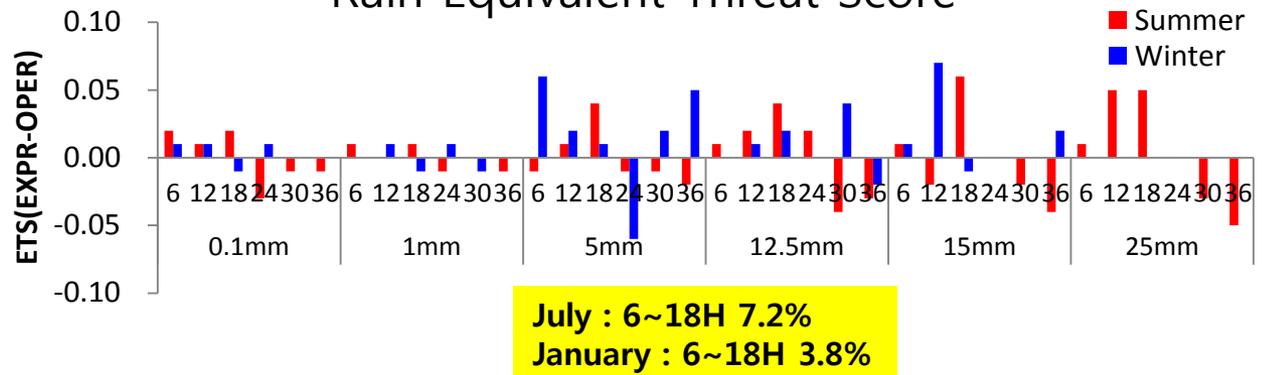
- Comparing the O-A, the change is biggest at CH3 after assimilating AMSU-B.
- It influences strongly on humidity and GPH fields near 500 hPa.

Ground-based GNSS in local model

40 domestic sites will be used operationally in next local model



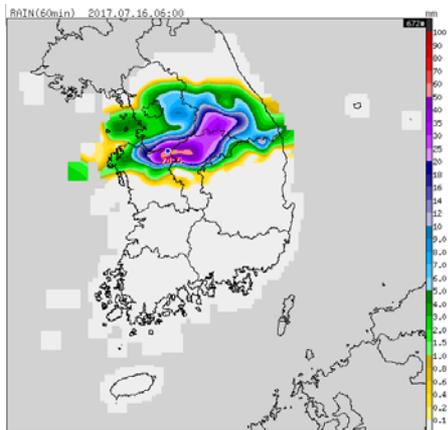
Rain Equivalent Threat Score



Eun-Hee Kim

Model Impact Result-case study

20170715 21UTC AWS

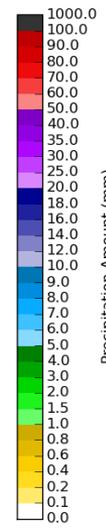
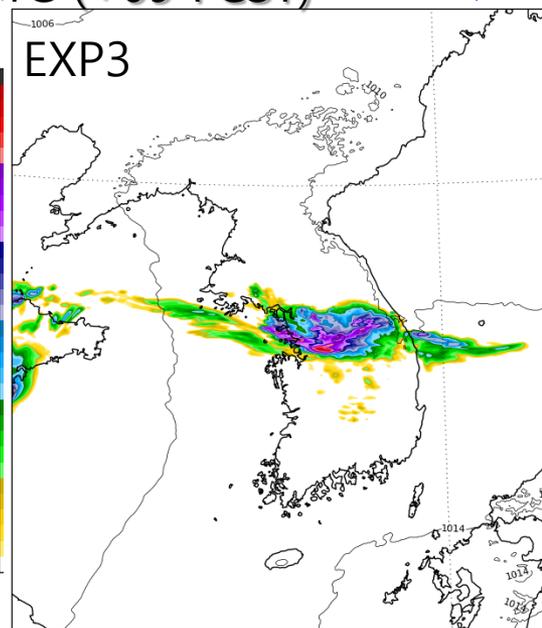
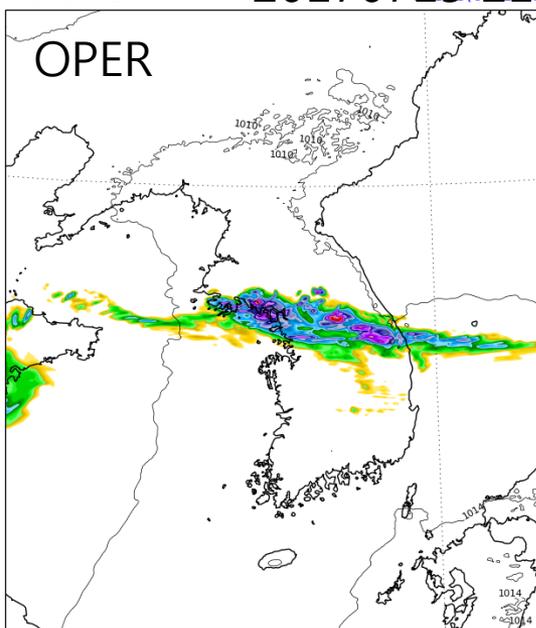


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MSLP&1H-PRECIP.

XLDPS(UM 1.5km L70)

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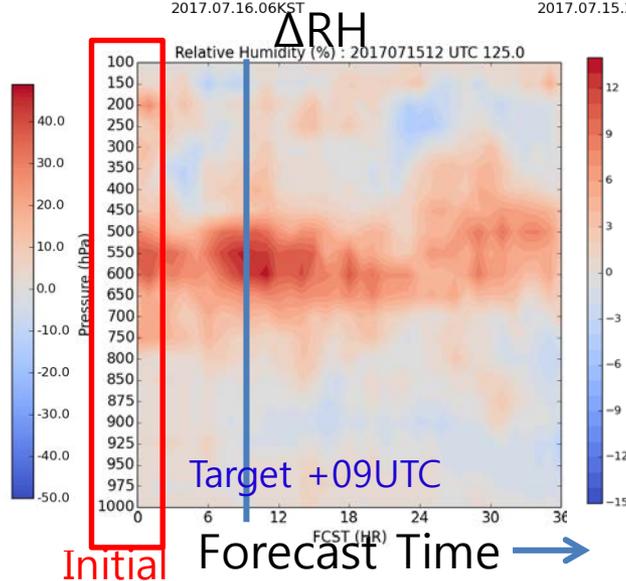
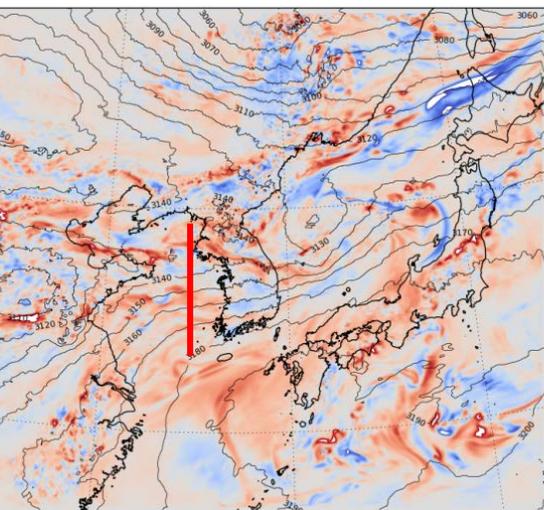
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2017.07.16.06KST

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2017.07.15.21KST

VALID TIME : 2017.07.15.21UTC (+09h FCST)
2017.07.16.06KST

TIME : 2017.07.15.12UTC
2017.07.15.21KST

Initial ΔRH 700 hPa



- When assimilating AMSU-B and groundGNSS additionally, these add humid in troposphere.
- Changes in the upper layer RH fields could affect the subsequent precipitation forecasts.

Summary and Future Plan

- KMA has been dedicating to assimilate satellite data to **fills the gap at the data sparse area** and provide **the moisture information**.
- By removing the cloud-affected radiances, the humidity and GPH field at 850~400hPa show better forecast performance.
- The skill score of precipitation forecast is improved but it is tend to overestimate the strong rainfall at the end of forecast.
- **Observation error and new bias correction method**
 - Additional polar-orbiting satellite data - IASI, ATMS, GPSRO...

Thanks for attention.

