

## 1. Objective

- Regional numerical weather prediction (NWP) model can do high resolution forecast. Forecast performance is effected by initial and boundary value.
- Data assimilation (DA) can improve initial condition by observations.
- A new AIRS single Field-of-View (SFOV) product retrieved by 1D-Var have higher spatial resolution than AMSU standard retrieval product.
- Use AMSU and AIRS SFOV soundings to evaluate the performance for introducing these two different data sets.

## 2. AMSU v.s. AIRS SFOV Retrievals

Table 1. Deference between AMSU and AIRS SFOV.

	AMSU	AIRS SFOV
Instrument	AMSU (MW) and AIRS (IR)	AIRS (IR) only
Horizontal resolution	45 km @ nadir	13.5 km @ nadir
Vertical resolution	T=28 levels; Q=15 levels	T/Q=101 levels
Cloud limit	0% ~ 80%	0% ~ 1%

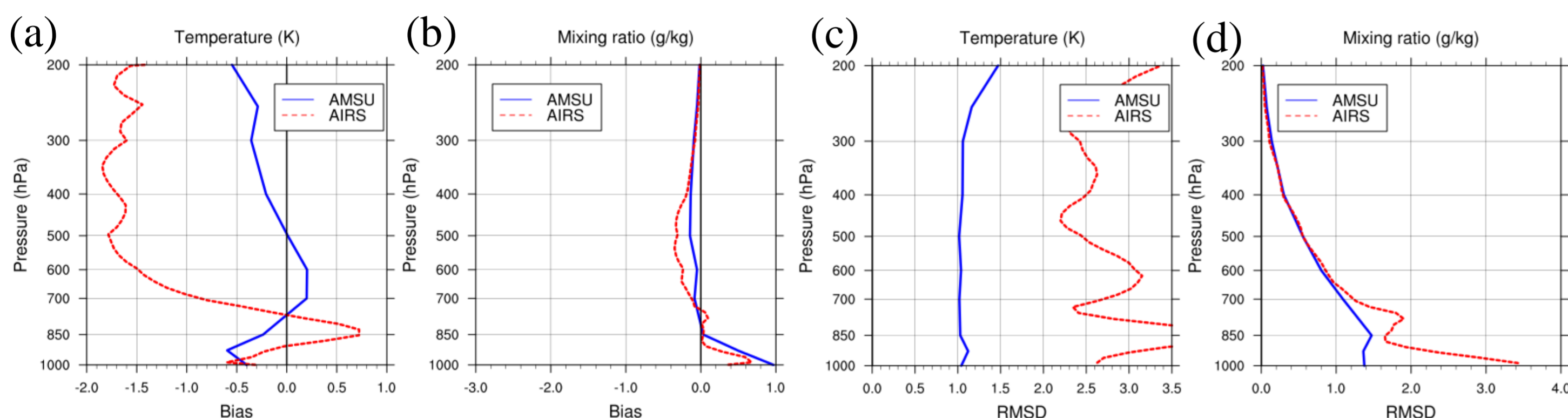


Figure 1. AMSU (AIRS SFOV) temperature (moisture) retrieval product's bias and RMSD in May 2012. ECMWF-interim reanalysis (0.25°x0.25°) data set is selected as reference.

## 3. Model and Experiment

### Weather Research and Forecasting Model V3.1.1

Initial time: 2012/06/08~13 (6 days) 0000 UTC  
 Initial: NCEP GFS ANL  
 Lateral boundary: ANL (DA) and FCST (FCST)  
 Domain: 3 one-way nested domains  
 Resolution: vertical 31 levels, horizontal 45/15/5 km

### WRF Data Assimilation v3.1.1

3D-Var  
 Assimilation interval: 6 h  
 Time windows: ±3 hr  
 Cycling period: 2 days

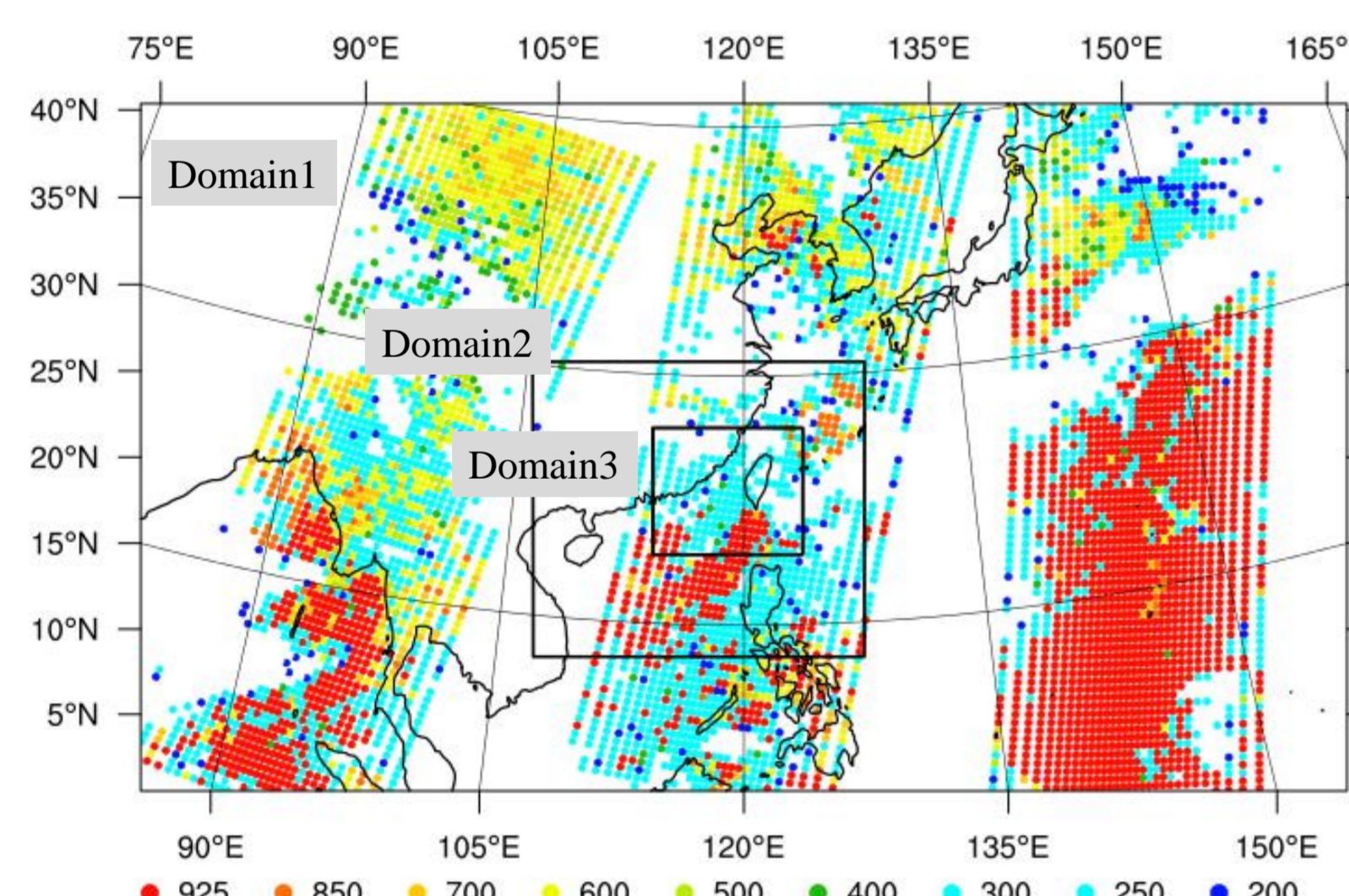


Figure 2. Domain and satellite observation. Point is a representative level from AMSU/AIRS data, all data above (include) this level have best quality.

### Table 2. Physics schemes in the model.

Physics scheme	
Microphysics	WSM 5-class
Longwave radiation	rrtm
Shortwave radiation	Dudhia
Boundary layer	Yonsei University
Cumulus	new Kain-Fritsch

### Table 3. Control and Experiments.

Experiment	Data Assimilated
CTRL	GTS
E_AMSU	GTS + AMSU
E_AIRS	GTS + AIRS SFOV

## 4. Preliminary Result

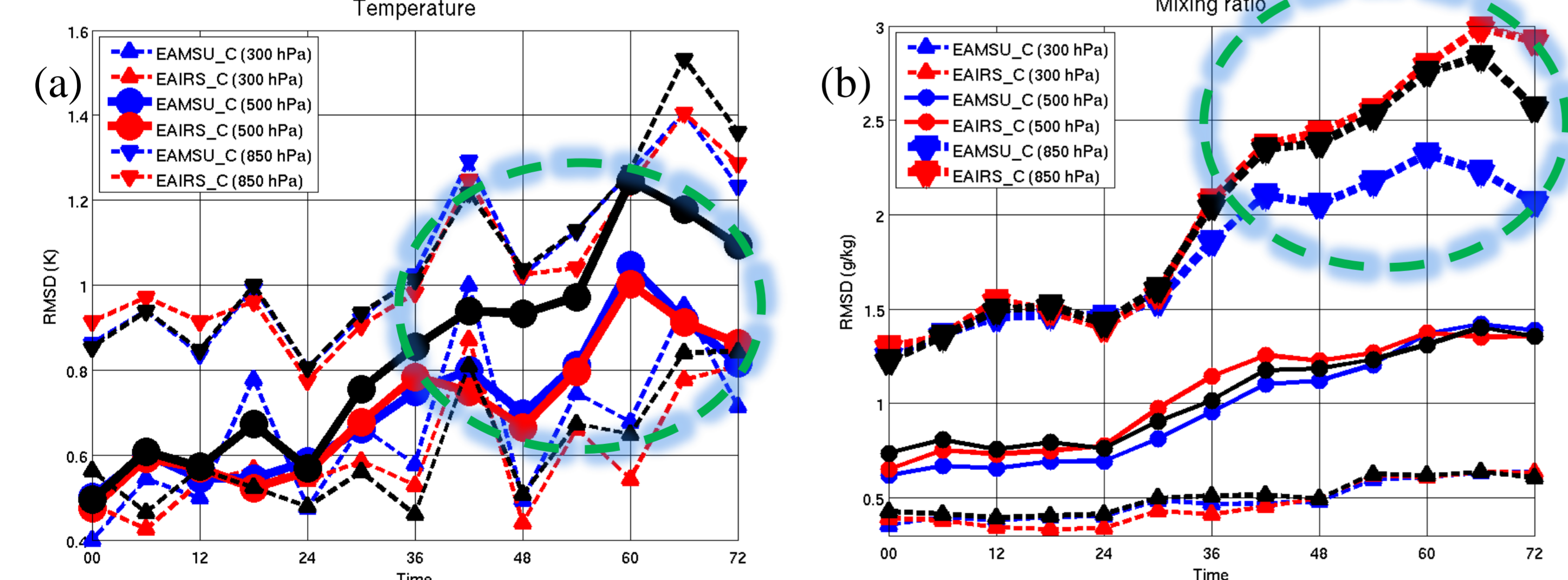


Figure 3. Control and Experiment run's forecast RMSD verification. (a) is Temperature and (b) is mixing ratio. Black lines are CTRL. A-axis is forecast time averaged by 6 runs in each experiment. ECMWF-interim reanalysis data set is selected as reference. Green circle represent lager difference, it will discussed in chapter 6.

- E\_AMSU better than E\_AIRS.
- There are many difference between AMSU and AIRS (ex. coverage, resolution, accuracy and observation/retrieval characteristic)
- Desiring new experiment to investigation the impact from these two retrieval product.

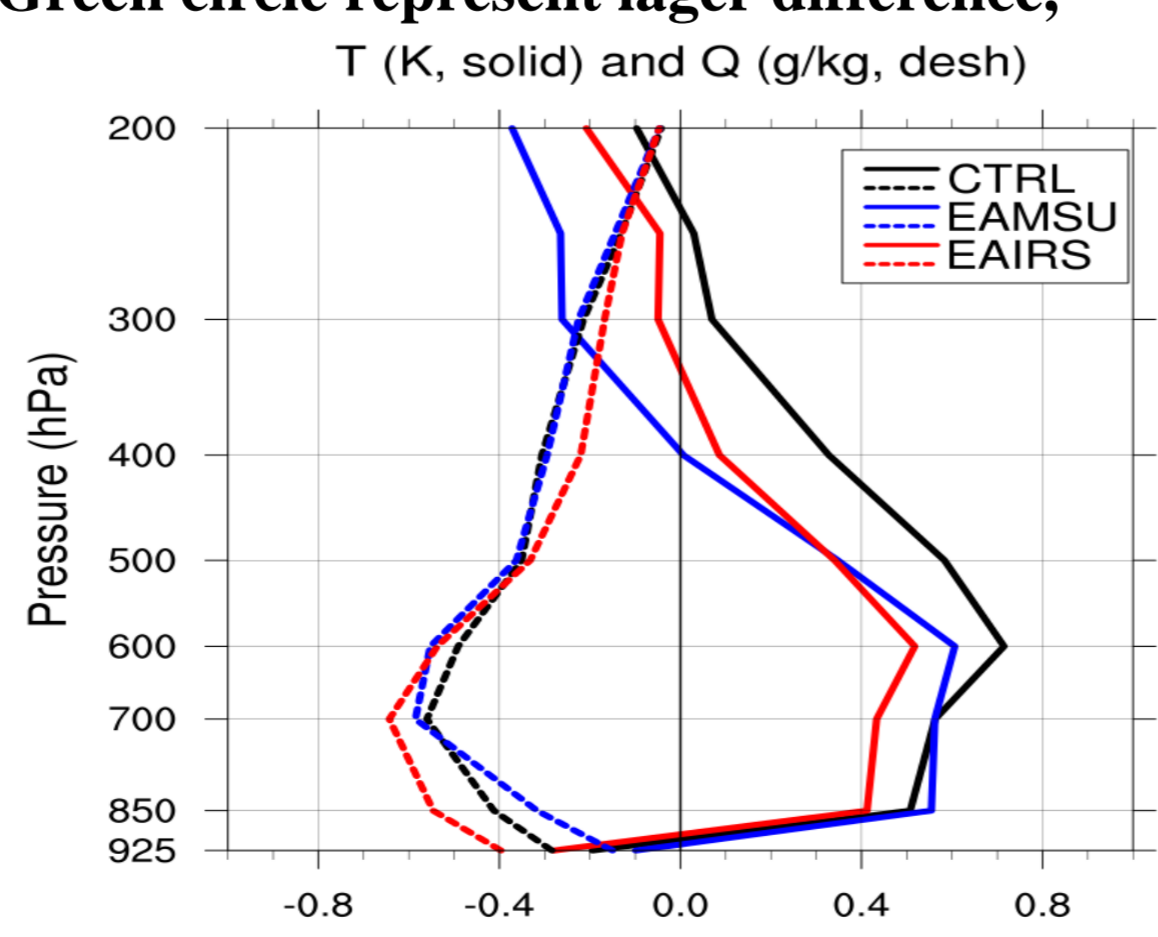


Figure 4. Mean bias form 0-72 forecast.

## 5. Data Spatial Quality Control-Collocation

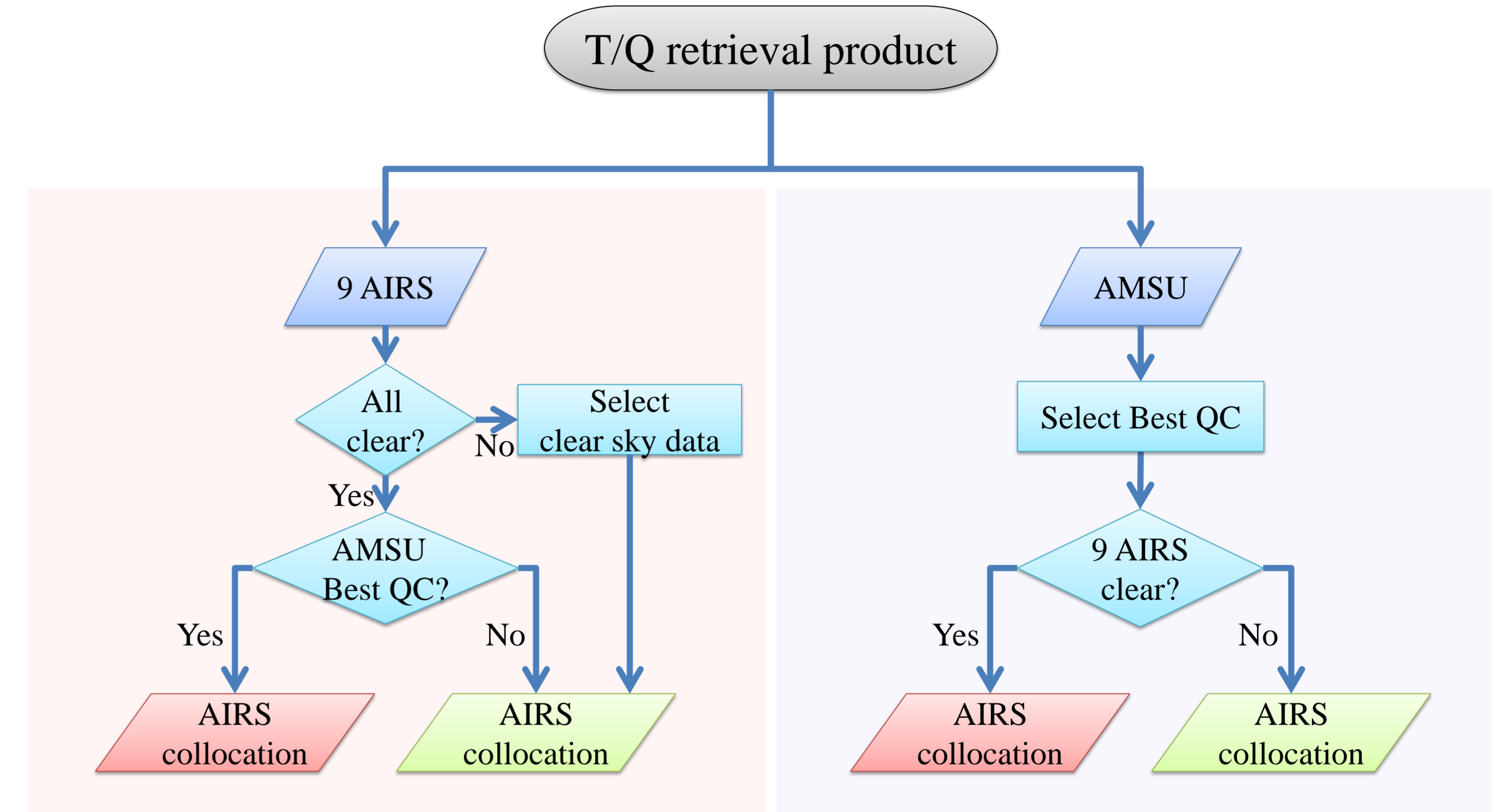


Figure 5. Collocation flow chart. AMSU's QC is base on TAirStd\_QC and H2OMMRLevStd\_QC provided from AMSU data set.

## 6. Sensitive Experiment

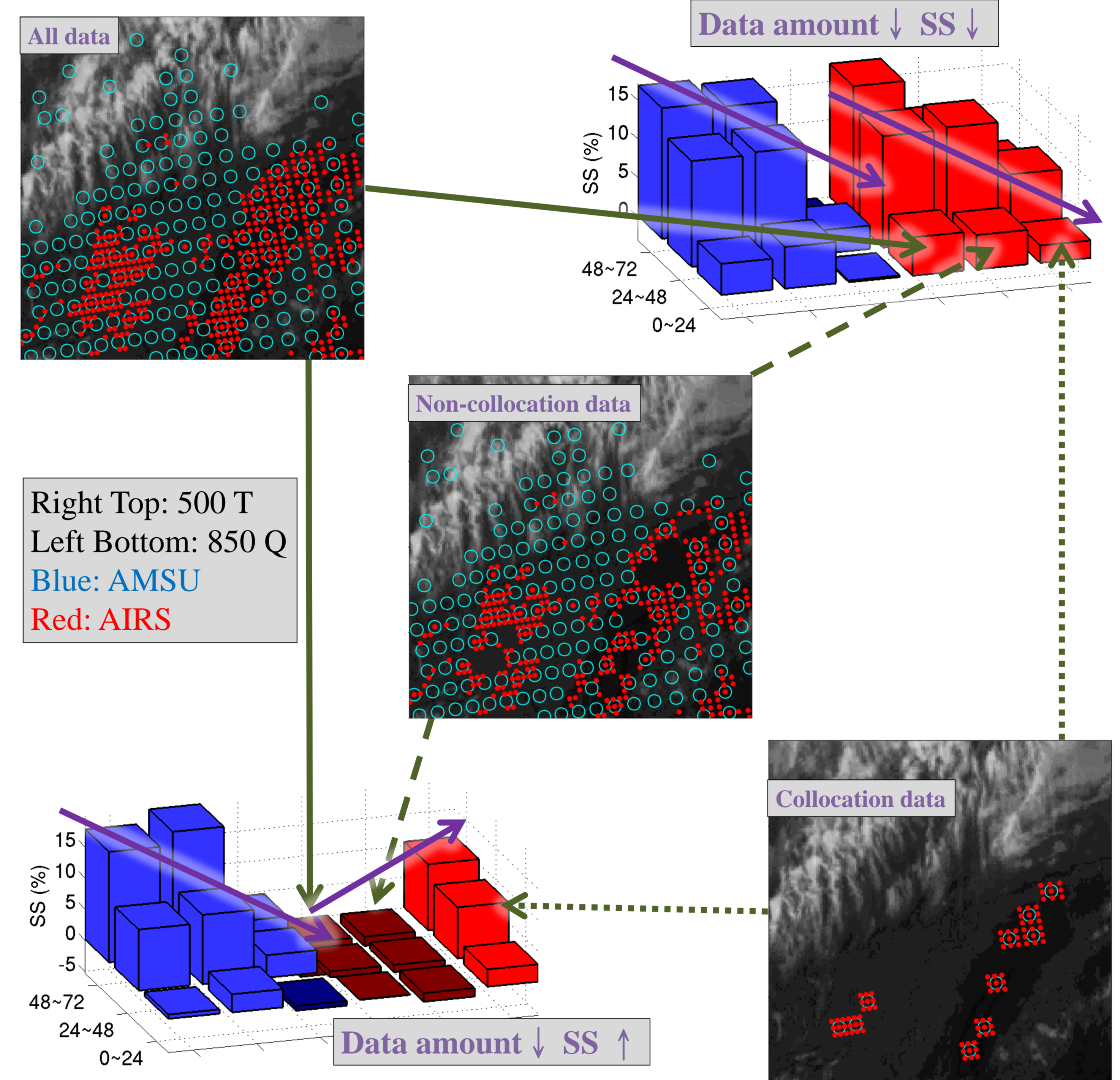


Figure 6. 500 hPa Temperature and 850 hPa mixing ratio's Skill Score in each experiment. Light: SS>0; dark: SS<0. SS represent improvement compared with CTRL.

$$\text{Skill Score}(SS) = \left( \frac{RMSD_{CTRL} - RMSD_{EXP}}{RMSD_{CTRL}} \right) * 100\%$$

- AMSU experiment: AMSU provide a stable retrieval product because it's observation/retrieval characteristic. It may cause this linear feature in SS/data.
- AIRS experiment: Mixing ratio have better performance than the other two. It may due to non-collocation data (green point shown in Fig.6). Data assimilation system may destroy existing system by taken and transmitted these non-cloud feature to entire region from non-collocation data.

## 7. Summary and Future Work

- If there have same data coverage between AIRS and AMSU, AIRS will have better positive impact in NWP forecast system than AMSU.
- Might be able to improve AIRS SFOV non-collocation problem in data assimilation system by using AIRS SFOV cloud retrieval product.
- Reducing observation influence radius or using another data assimilation method which can transmitting data information better may improve high resolution satellite retrieval product application in NWP forecast.

## References

Liu, C.-Y.\*, G.-R. Liu, T.-H. Lin, C.-C. Liu, H. Ren, and C.-C. Young (2014), Using Surface Stations to Improve Sounding Retrievals from Hyperspectral Infrared Instruments, *IEEE TRANS. ON GEOSCIENCE AND REMOTE SENSING* (accepted).  
 Liu, C.-Y.\*, J. Li, P. Zhang, T. J. Schmit (2012), Applications of Full Spatial Resolution Space-Based Advanced Infrared Soundings in the Preconvection Environment, *WEATHER AND FORECASTING*, 27, 515-524, doi:10.1175/WAF-D-10-05057.1.