# **Development of KIAPS Observation Processing System :** (1) AMSU-A Bias Correction Modules

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### Introduction

#### **Bias and Data Assimilation**

All data assimilation systems are affected by biases, caused by problems with the data, by approximations in the observation operators used to simulate the data, by limitations of the assimilating model, or by the assimilation methodology itself. A clear symptom of bias in the assimilation is the presence of systematic features in the analysis increments (Dee, 2005).

### **Objective**

To introduce the AMSU-A radiance pre-processing and bias correction at the KIAPS Observation Processing System (KOPS)

|  |  | Monitoring :                                    |
|--|--|---|
| AMSU-A ch05: Obs TB                                | AMSU-A ch05: Obs TB (SD)                           | wonitoring :                                    |
| 90°N<br>60°N<br>30°N<br>0°                         | 90°N min= 0 max= 6 mean= 3.782 stdy= 0.6827        | BUFR extracted                                  |
| 30°S   | 30'5   | • Observed TB (0.35° x 0.23°)                   |
| 60°5<br>90°5<br>180° 120°W 60°W 0° 60°E 120°E 180° | 60°5<br>90°5<br>180° 120°W 60°W 0° 60°E 120°E 180° | In channel 5, observed TB                       |
|  |  | low latitude, but it decrease                   |
| 215 220 225 230 235 240 245 250 255 260 265        | 0.0 0.6 1.2 1.8 2.4 3.0 3.6 4.2 4.8 5.4 6.0        | latitude. The land variab                       |
| AMSULA ch05: Rkg TR                                | AMSULA ch05' Rkg TR (SD)                           | standard deviation) is more th                  |
| 90'N min= 221 max= 263.1 mean= 246 stdv= 8.928     | min= 0 max= 5.998 mean= 3.595 stdv= 0.611          | Standard deviation) is more th                  |
| 60'N   | 60'N   |   |
| 30'N   | 30°N   | <ul> <li>Background TB (0.35° x 0.23</li> </ul> |
| or the second a second                             | · · · · · · · · · · · · · · · · · · ·              | Monthly mean of backgroun                       |
| 30°5   | 30'5   |   |
|  |  | Madal autout: a a away00                        |

## **Results and Discussions**



180° 120°W 60°W 0° 60°E 120°E







AMSU-A ch06: O-B after airmass correction AMSU-A ch06: O-B after airmass correction (S min= -0.9994 max= 0.9108 mean= -0.001468 stdv= 0.04717 min= 0 max= 0.9984 mean= 0.1658 stdv= 0.0522

### Monitoring : **Bias corrected innovation**





After airmass bias correction, some regions are affected by atmospheric variability due to temperature variations.

Channels 06~11

Airmass bias is successfully controlled by multiple linear regression coefficients and off-set.



~0.3 K.















(1) Scan Bias Correction

• **Goal :** To remove the systematic errors attributed to the observations, the radiative transfer model and pre-processing steps

**d** : radiance difference Ø : latitude band ✓ Harris and Kelly (2001) :  $d(\emptyset, \theta) = O_j(\emptyset, \theta) - O_j(\emptyset, \theta=0)$ **θ** : scan angle **O** : observation (j = channel) ✓ Weng et al. (2012) :  $(O - \mu^{\circ}) - (B - \mu^{b}) = O - B - (\mu^{\circ} + \mu^{b})$ **B**: background **µ°**: observation bias µ<sup>b</sup>: model bias

#### (2) Airmass Bias Correction

• **Goal** : To remove the systematic errors attributed to the different thermodynamics properties scanned atmosphere and the surface





Algorithm : global linear regression of the scan-corrected innovations against two predictors (850-300 hPa and 200-50 hPa thicknesses) to correct the airmass bias

**B** : airmass bias (j = channel) ✓ Harris and Kelly (2001) :  $B_j = \sum_{j=1}^{11} A_{ji} X_i + C_j$ A, C: coefficients by linear regression X<sub>i</sub>: predictors (i = number of predictor)

Model based predictors

- Thickness 850-300 hPa
- Thickness 200-50 hPa
- Surface skin temperature
- Total column water vapor

✓ Weng et al. (2012) : O-B statistics

> Multiple linear regression :  $Y_i = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \ldots + \beta_k X_{ki} + \varepsilon_i$ ,  $i = 1, 2, \ldots, n$ 



### Summary

• We have developed the modules of satellite radiance data pre-processing and quality control at KOPS, which include observation operators to interpolate model state variables into radiances in observation space. AMSU-A level-1d radiance data was extracted using the BUFR (Binary Universal Form for the Representation of meteorological data) decoder and a first guess was calculated with RTTOV10.2.

• For initial quality checks, (1) cloud contaminated data were removed for use of clear-sky radiance data and (2) AMSU-A channels for assimilation, rejection, or monitoring were respectively selected for surface type and topography since the errors from the skin temperature were caused by inaccurate surface emissivity.

• KOPS AMSU-A bias correction modules have developed in two steps based on 30-day innovation statistics. The scan bias correction coefficient was calculated individually for each channel, satellite, and scan position. Then a global multiple linear regression of the scan-corrected innovations against several predictors (e.g., 850-300 and 200-50 hPa thicknesses) was employed to correct the airmass bias.

• It appears to be successful in controlling the scan and airmass bias in the crucial channels which sound tropospheric and stratospheric temperature (just below 50 km altitude).

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