



# **Cross-track Infrared Sounder Spectral Gap Filling toward Improving Inter-calibration Accuracy**

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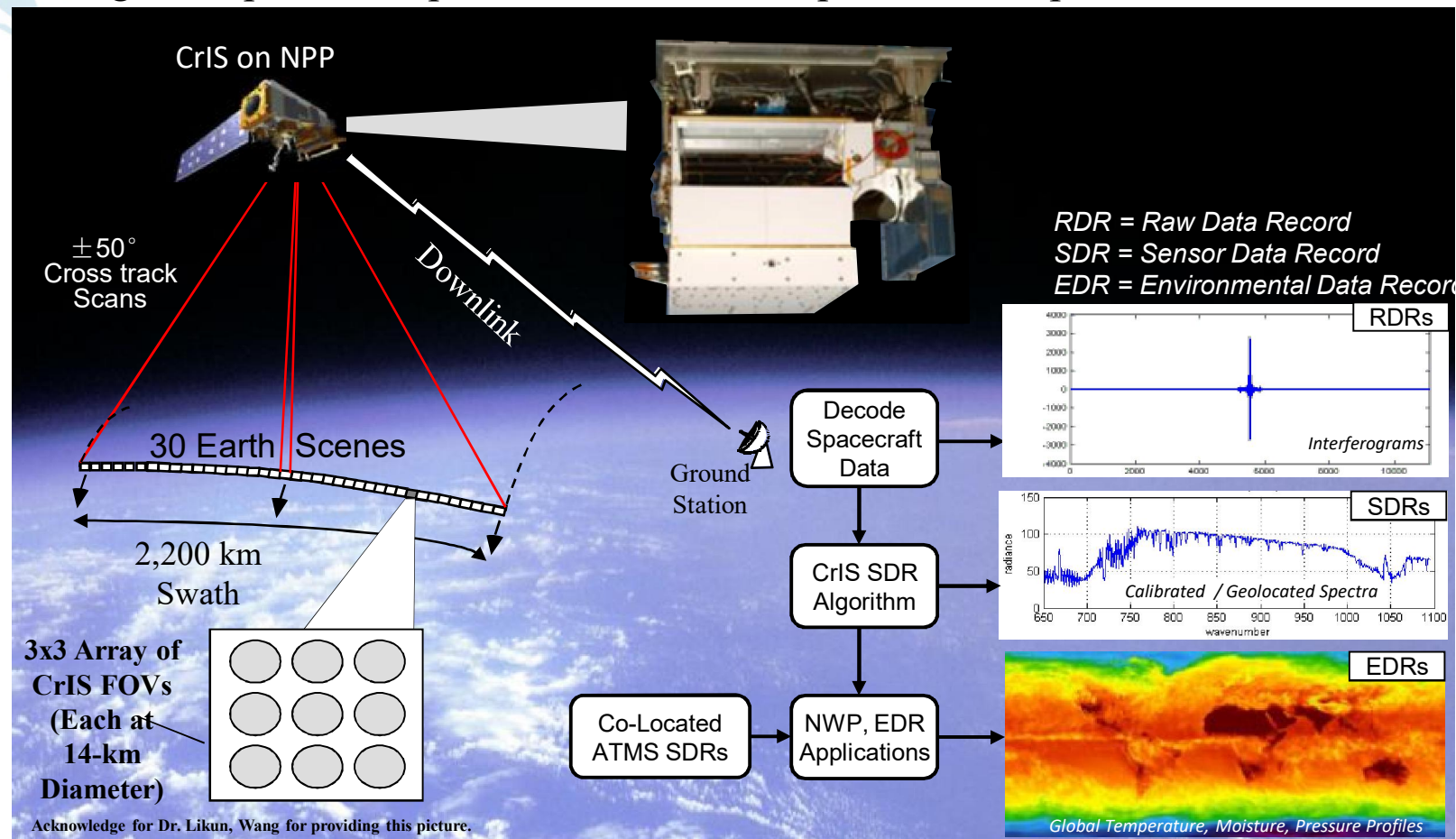
(CICS/ESSIC/Univ. of Maryland, College Park, MD)

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## Background



❖ *CrIS (Cross-track Infrared Sounder)* is a hyperspectral infrared sounder mainly for providing atmospheric temperature and moisture profiles to improve weather forecast.



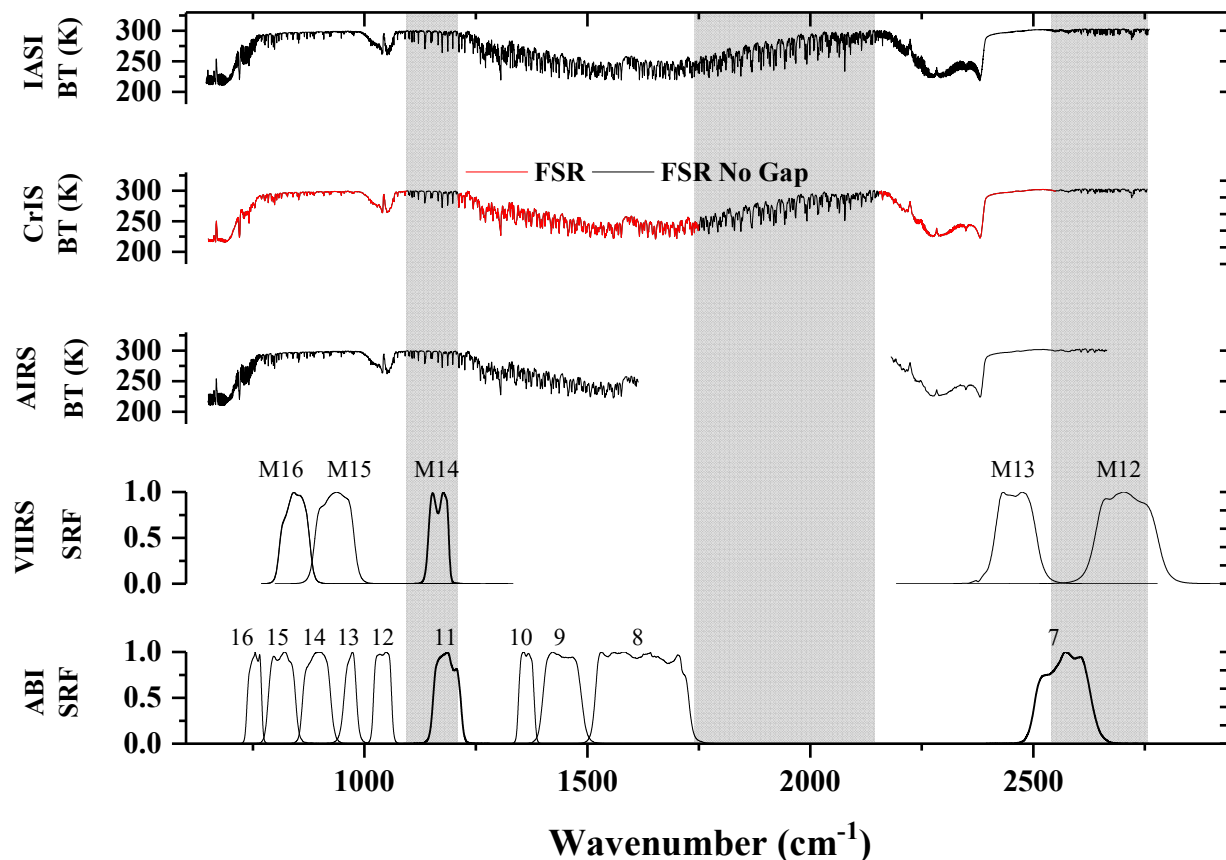
It is a Michelson interferometer with 2211 (Full Resolution,  $0.625 \text{ cm}^{-1}$ ) channels over three wavelength ranges: 2 LW infrared ( $650\text{-}1095 \text{ cm}^{-1}$ ), MW infrared ( $1210\text{-}1750 \text{ cm}^{-1}$ ), and SW infrared ( $2155\text{-}2550 \text{ cm}^{-1}$ ).

## Background



Due to its excellent performance (*high radiometric and spectra accuracy with low noise*), CrIS radiance is also used as an *infrared reference* to check calibration accuracy of broadband sensors, such as AVHRR/ABI on GOES, VIIRS on NPP.

### Instruments Spectral Coverage



### CrIS AS A REFERENCE

$R_{\text{simulated-broadband}}$

$$= \frac{\int_{v1}^{v2} R_{CrIS}(v) \cdot S(v) \cdot dv}{\int_{v1}^{v2} S(v) \cdot dv}$$

### Spectral Gaps

impact the accuracy of  
inter-comparison between  
CrIS and other sensors !!!



# Outline



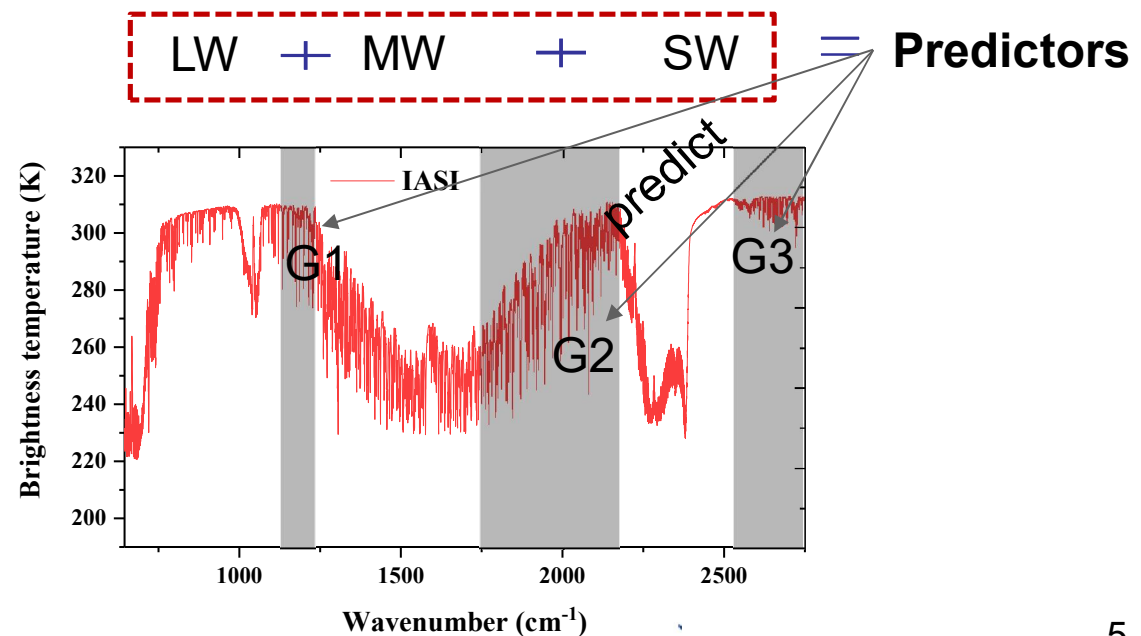
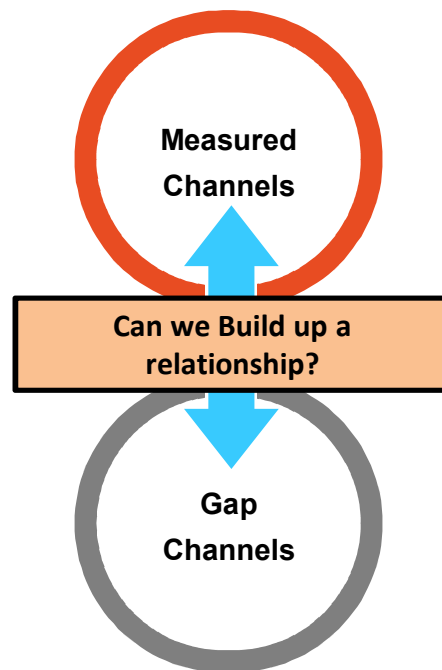
- **Methodology of the CrIS gap channels prediction**
  - Training dataset generation
  - Principal Component Regression based spectral gap prediction
- **Prediction results validation**
  - Theoretical accuracy analysis
  - Inter-comparison with hyper-spectral infrared sounders
- **Application**
  - CrIS – VIIRS inter-comparison on M14
  - CrIS – ABI inter-comparison on channel 7, 8 and 11
- **Conclusions and future work**

# Methodology



## The theoretical basis of this method is that

- the top of atmosphere (TOA) wavenumber dependent radiances are highly correlated with each other. (the CrIS gap channels spectral information possibly has already existed in the measured channels)



# Methodology

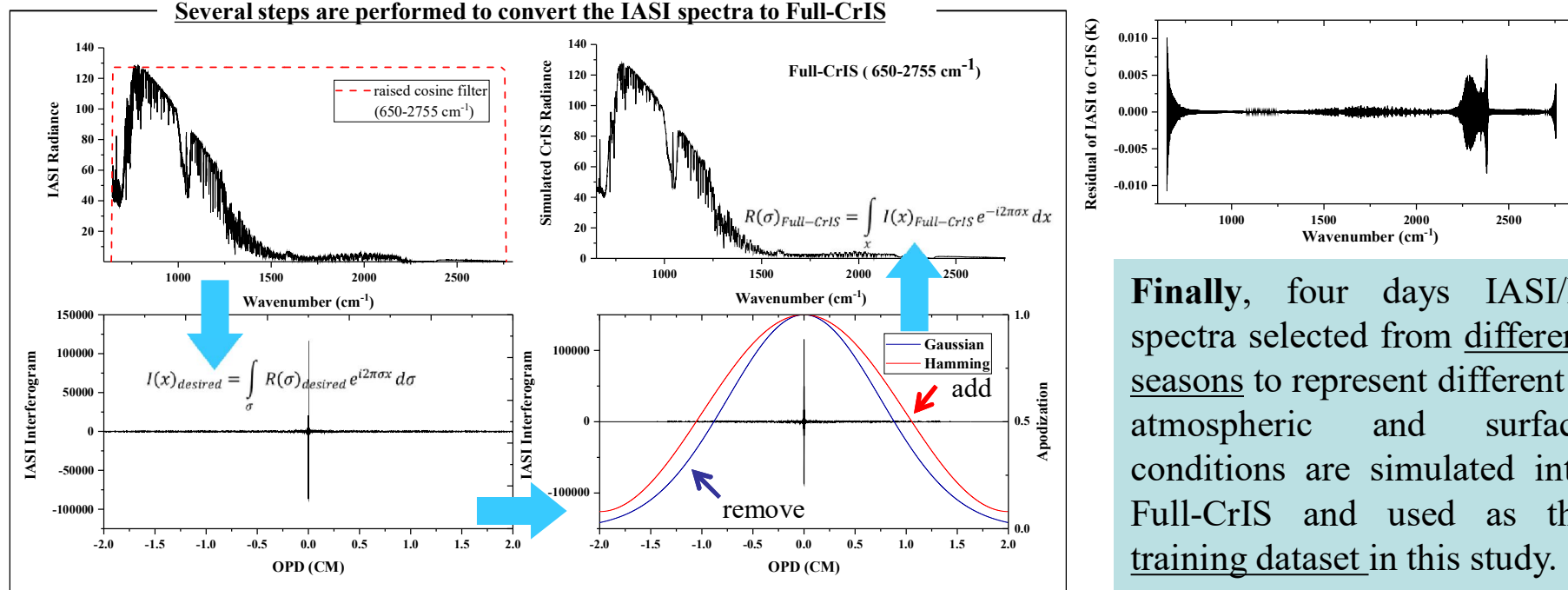
## Training dataset generation

To establish this relationship, a **Full-CrIS** training dataset spectra which includes both measured and gap channels needs to be built up at first.

•The Full-CrIS training dataset was simulated from spectra measured by **Infrared Atmospheric Sounder Interferometer (IASI)**.

- ❑ IASI with 8461 spectral channels sampled every  $0.25 \text{ cm}^{-1}$ , covering 'ALL' the CrIS spectral information
- ❑ The conversion residual from IASI to CrIS spectrum is very small (basically less  $0.01 \text{ K}$ )

Several steps are performed to convert the IASI spectra to Full-CrIS



Finally, four days IASI/B spectra selected from different seasons to represent different atmospheric and surface conditions are simulated into Full-CrIS and used as the training dataset in this study.

# Methodology

## Principal Component Regression (PCR) based spectral gap prediction

Suppose the gap channel radiances can be predicted by using

$$R_{Gap}(\sigma) = \sum_{i=0}^k R_{MEA}(i) \cdot P(\sigma, i) + \beta(\sigma)$$

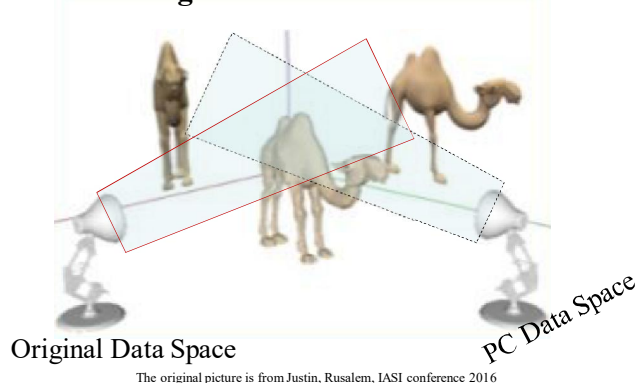
Gap channel radiance
Measured channel radiances
Prediction Coefficients

This part can not be directly derive!

**However**, it is hard to straightforward derive above **Prediction Coefficients** by using a fitting method. This is because the wavenumber dependent radiances are highly correlated with each other (**Multi-collinear**).

## Principle Component Analysis (PCA)

PCA is orthogonal transformation technic



### PCA

- ☐ Reduce the dimension of the dataset
- ☐ Conserve a maximum of the effective information
- ☐ Denoising the random noise of the dataset
- ☐ Decorrelation of the interdependent variables

Therefore, a **Principle Components based Regression (PCR)** method is used in this study to derive the prediction coefficients.



# Methodology

## Principal Component Regression (PCR) based spectral gap prediction

If the CrIS gap channel radiance  $y \in \{y_1, y_2, \dots, y_m, m = 1158\}$  can be predicted by the measured channel radiances  $x \in \{x_1, x_2, \dots, x_n, n = 2211\}$  using Eq.1,

$$y_i = \beta_{0,i} + \beta_{1,i} \cdot x_1 + \beta_{2,i} \cdot x_2 + \dots \beta_{j,i} \cdot x_j + \dots + \beta_{n,i} \cdot x_n + u_i \quad (1)$$

where  $\beta_{j,i}$  is the prediction coefficients,  $u_i$  is the residual,  $j$  is the  $j$ th measured channel radiance and  $i$  is the  $i$ th gap channel. Then, in order to derive the above the coefficients  $\beta$  and  $u$ , all the gap channel radiances with a total of  $t$  observations can be written in the matrix form,

$$Y = D_X \times B + U \quad (2)$$

where  $Y$  is an  $m \times t$  matrix containing the CrIS gap channel radiances ( $m$  channels and  $t$  samples,  $D_X$  is the design matrix with a dimension  $(n + 1) \times t$  with the first column elements set to 1 and the rest of columns containing the CrIS measured channel radiances  $X$  with a dimension  $n \times t$ ,  $B$  is the prediction coefficient matrix with a dimension  $m \times (n + 1)$ , and  $U$  is the  $m \times n$  residual matrix representing the matrix manipulation.

The prediction coefficient matrix  $B$  then can be calculated through Eq.3 with the **Full-CrIS** training dataset,

$$B = (D_X^T \times D_X)^{-1} \times D_X^T \times Y \quad (3)$$

where  $T$  and the superscript -1 represent the transpose and inverse, respectively.



# Methodology

## Principal Component Regression (PCR) based spectral gap prediction

To solve the multi-collinear problem, a principal component transformation is performed on  $X$  to de-correlate the measured channel radiances into orthogonal **principal component scores**  $pcs_X$  before deriving the coefficients.

by replacing  $X$  in the design matrix with  $pcs_X$ , the relationship between the principal component scores of the measured channel radiances and the gap channel radiances now can be properly established through Eq.4 with the training dataset,

$$B_{pcs} = (D_{pcs_X}^T \times D_{pcs_X})^{-1} \times D_{pcs_X}^T \times Y \quad (4)$$

where the  $B_{pcs}$  is the principle components based gap channel prediction coefficient matrix.

### Parameter Optimization

- Since most of the Earth spectral variances are mainly distributed in the first few principal components, **only the leading  $k_X$  principal component scores are used as the predictors.**
- To improve the fitting accuracy, **the noise decreased gap channel radiances  $Y_{dec}$**  are calculated firstly. Here, the  $Y_{dec}$  are reconstructed from their leading  $k_Y$  principal component scores.

Eventually, based on the optimized predictors and response variables, the prediction coefficient matrix  $B_{pcs}$  can be successfully regressed with training dataset by using (5),

$$B_{pcs} = (D'_{pcs_X}^T \times D'_{pcs_X})^{-1} \times D'_{pcs_X}^T \times Y_{dec} \quad (5)$$

the symbol ' indicate the truncated matrix or its result is truncated

# Methodology

## Principal Component Regression (PCR) based spectral gap prediction

At last, the CrIS gap channel radiances in  $Y_{CrIS}$  are predicted from the real CrIS data  $X_{CrIS}$  through (6),

$$Y_{CrIS} = [(X_{CrIS} - \bar{X}) \times N_X^{-1} \times E_X^T]' \times P_{pcs} + C_{pcs}^T \quad (6)$$

where  $N_X$  and  $\bar{X}$  are the instrument noises and mean radiances derived from the **Full-CrIS** training dataset respectively;  $C_{pcs}$  is the constant row of the  $B_{pcs}$  and  $P_{pcs}$  is the coefficient rows of the  $B_{pcs}$ .

By combining all the above matrices together, Eq.6 finally becomes

$$Y_{CrIS} = X_{CrIS} \times P + C \quad (7)$$

- $P$  is the CrIS gap channel prediction coefficients
- $C$  is the corresponding gap channel constant

**PS :** The uncertainties caused by the difference of instrument noises between **real CrIS** and **Full-CrIS** should be neglected. This is because the CrIS instrument noises is overall similar to that of the Full-CrIS and the noises left in the predictors are basically very small after the PC rotation.

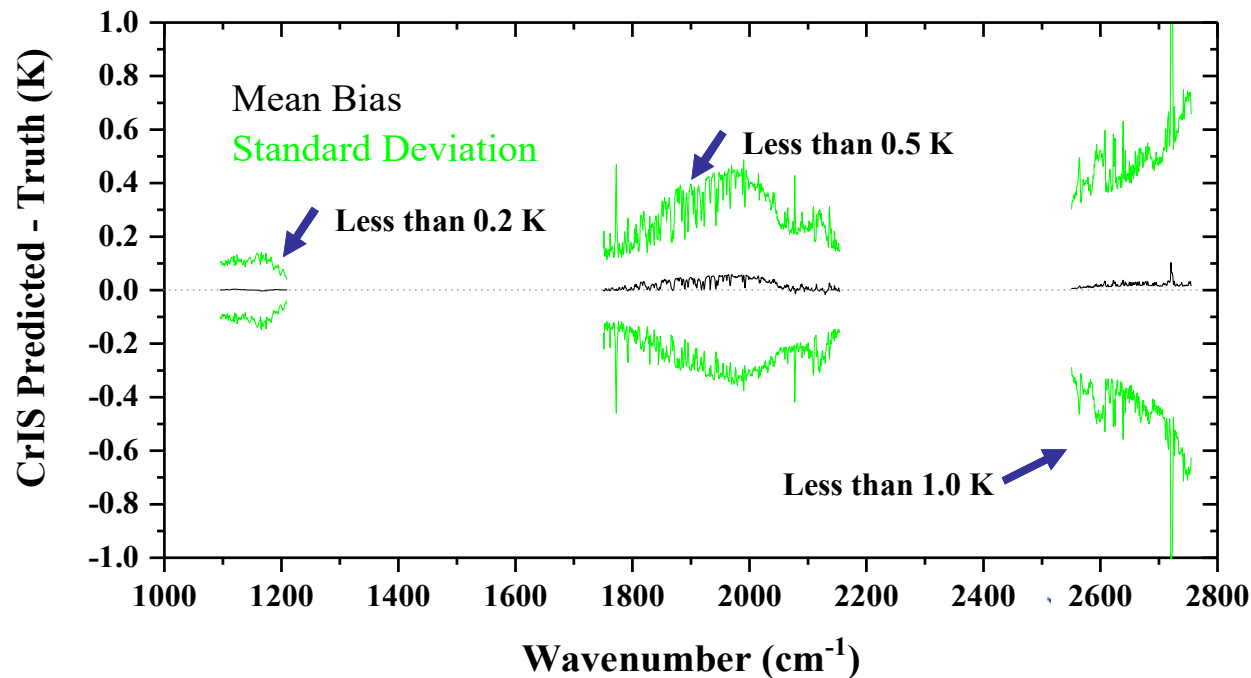
# Prediction results validation



## Theoretical accuracy analysis

The evaluation of the prediction results is not easy since we do not have real measured data in these CrIS gap channels.

- ❖ One-day IASI/B data which are different from the training dataset, are selected and converted into Full-CrIS spectra.
- ❖ The measured channels of the Full-CrIS are used as predictors, while the gap channels of the Full-CrIS are used as the truth to check the prediction accuracy.





# Prediction results validation



## CrIS – AIRS

**SNO selection:** FOV distance : less than 6.85 km ; Time difference: less than 2 (polar) and 15 minutes (tropical);

View angle difference:  $\text{abs}(\cos(\text{zen1}) - \cos(\text{zen2})) < 0.01$  ; VIIRS M16 :  $\text{std}(\text{M16}) / \text{mean}(\text{M16}) < 0.05$ ; NADIR: FOR 15 and 16

### SNO (Simultaneous Nadir Overpass ) for AIRS:

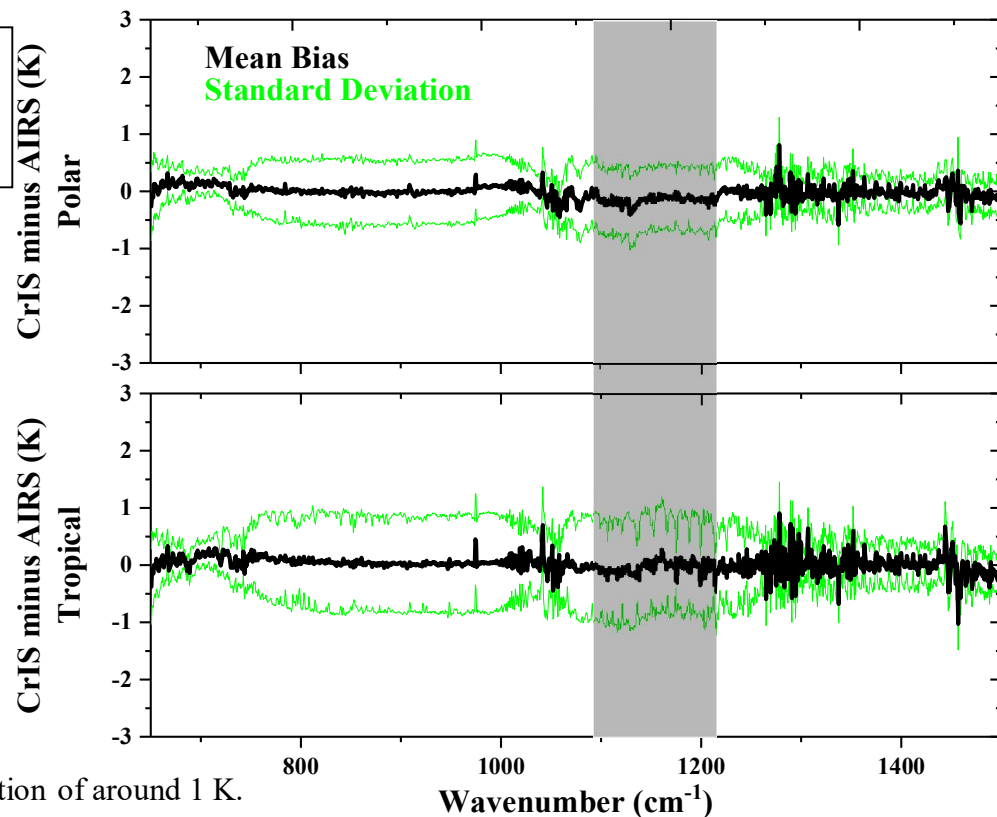
2016-11-10, 2016-11-13, 2016-11-16, 2016-11-18

**AIRS – CrIS conversion:** AIRS is a grating spectrometer, while CrIS is a interferometer. (UMBC, Howard E. Motteler 2016)

- deconvolve AIRS L1C (2465 channels) data to 0.1 cm<sup>-1</sup> resolution spectrum;
- convert it to desired CrIS band by using the same IASI-CrIS conversion method .

### SNO difference in LW gap channels:

Their BT differences are close to zero with a standard deviation of around 1 K.



As shown, the predicted CrIS gap spectra agree well with those observed by AIRS. <sup>12</sup>  
The BT differences over the predicted channels are similar to those of the measured channels.



# Prediction results validation



## CrIS – IASI

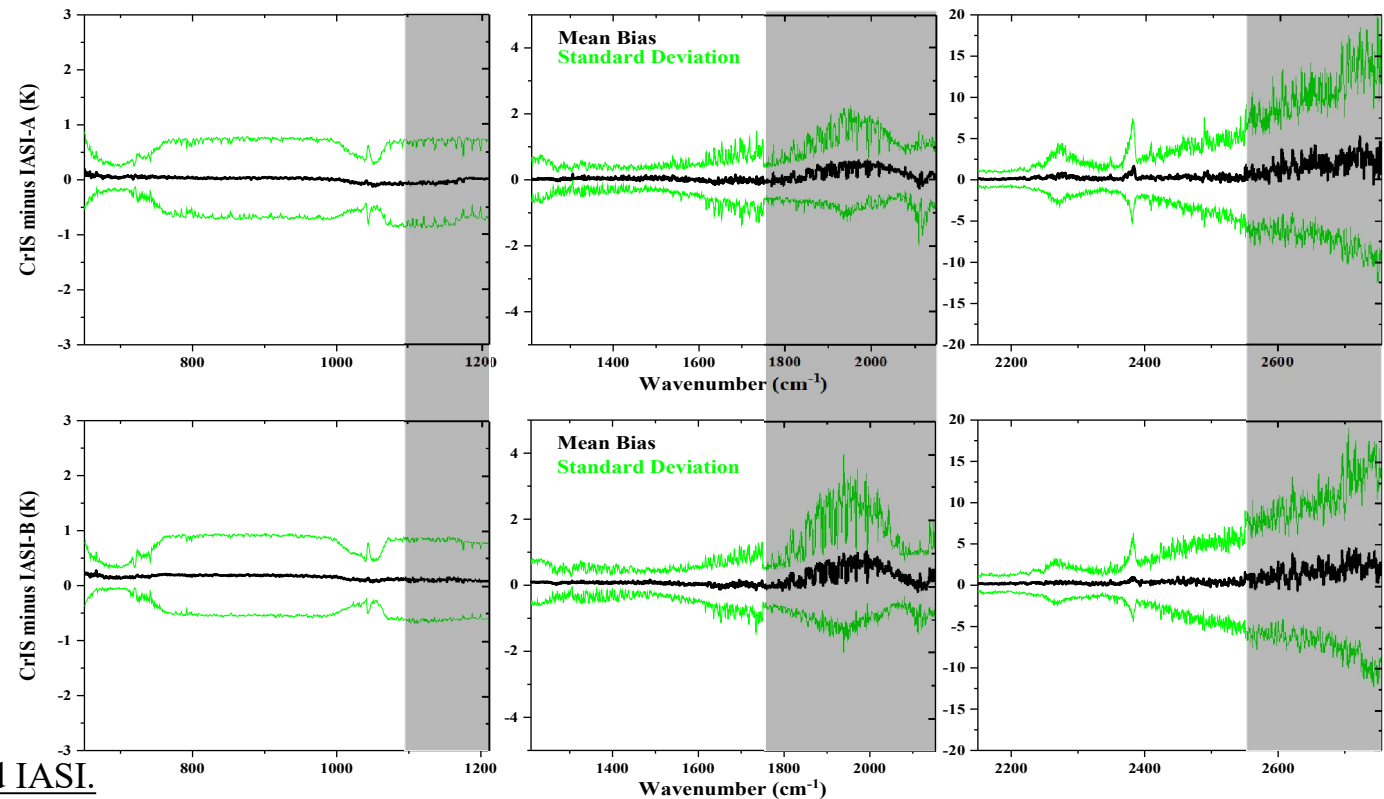
### SNO for IASI-A:

2016-08-13, 2016-08-14,  
2016-10-03, 2016-10-04

### SNO for IASI-B:

2016-09-09, 2016-09-10,  
2016-10-30, 2016-10-31

**SNO selection:** FOV distance : less than 6.5 km ; Time difference: less than 2 minutes ; NADIR: FOR 15 and 16  
**View angle difference:**  $\text{abs}(\cos(\text{zen1}) - \cos(\text{zen2})) < 0.01$  ; VIIRS M16 :  $\text{std}(\text{M16}) / \text{mean}(\text{M16}) < 0.05$



## Good Agreements

are observed between CrIS and IASI.

- Their spectral differences in LW are similar to that of AIRS-CrIS inter-comparison.
- Due to the SW radiances are much lower than LW given the same BT, the BT uncertainties are increase with the increase of wavenumber.
- In general, the predicted CrIS gap spectra agree well with those measured channels through the comparison with IASI, suggesting they may have potential ability to be used in inter-calibration.

# Application

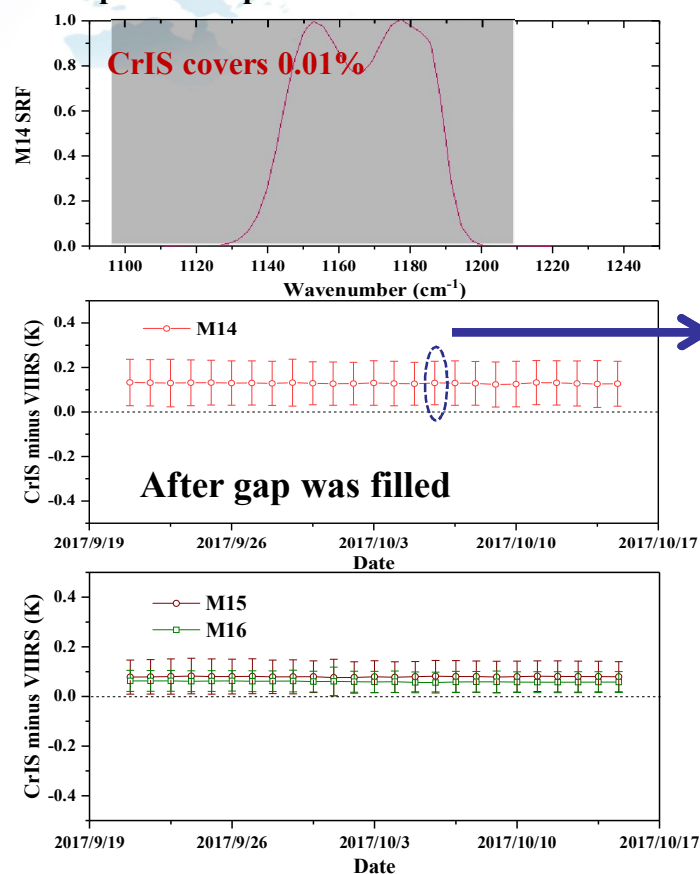


## CrIS - VIIRS

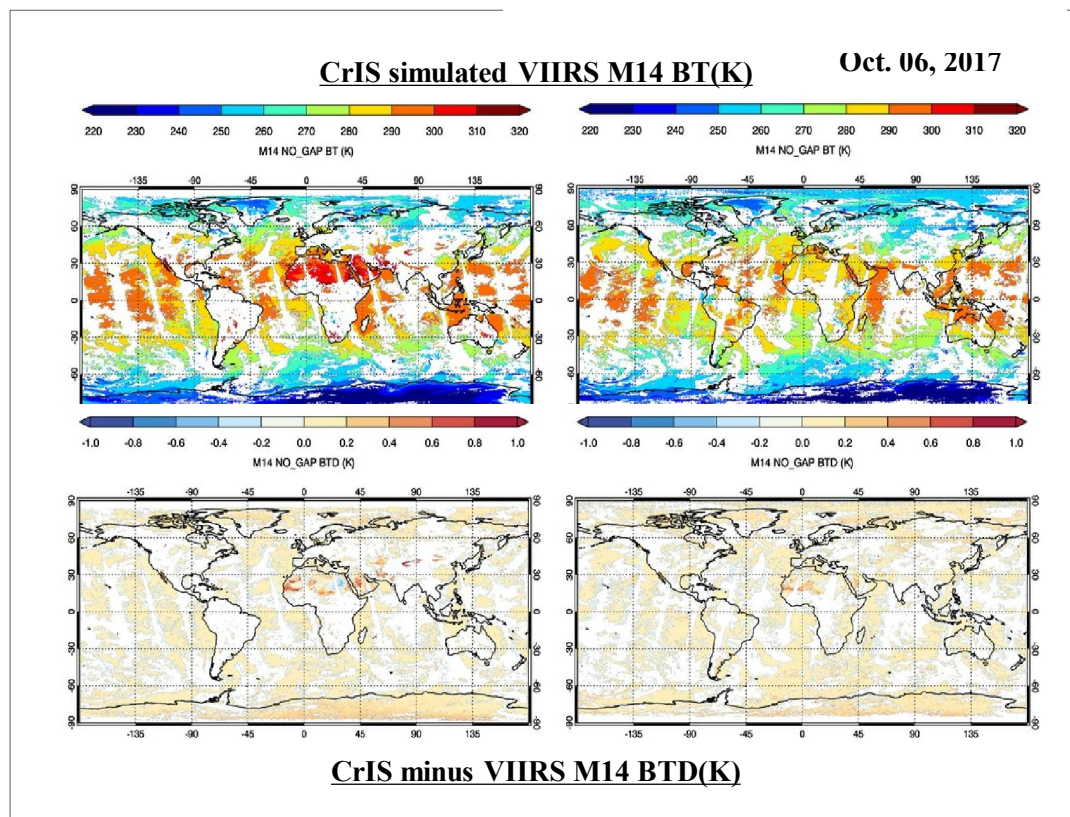
**SNO selection:** The CrIS and VIIRS are paired together by using a fast collocation method proposed by Wang et al. (2016).

**Only uniform scenes are selected:** VIIRS M16 :  $\text{std}(\text{M16}) / \text{mean}(\text{M16}) < 0.01$  ; VIIRS within CrIS are averaged.

Spectral response functions of VIIRS M14



## Global distribution



**Gap-filling method works very well and stable** in the M14 LW window channel, even though the bias and standard deviation are a little higher than M15 and M16.

# Application

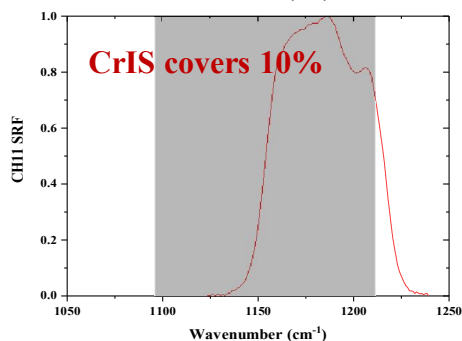
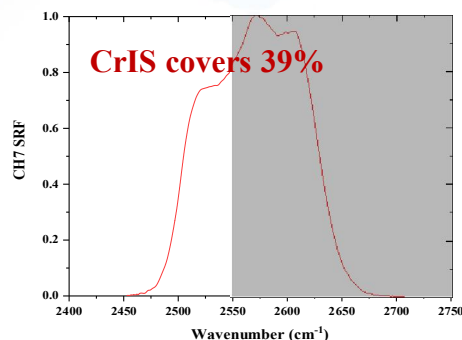


## CrIS - ABI

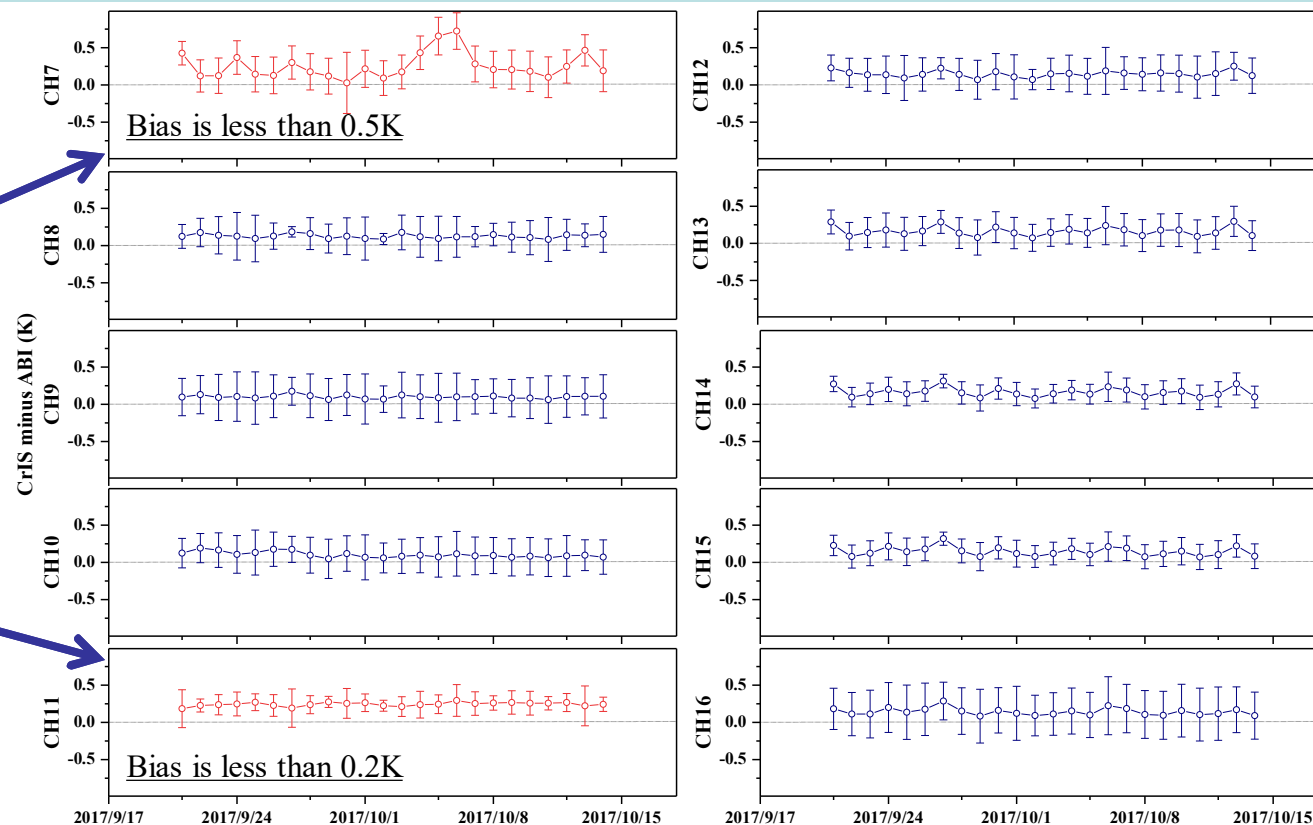
**SNO selection:** FOV distance : less than 7 km ; Time difference: less than 10 minutes ; Nadir: FOR 14, 15, 16 and 17

**View angle difference:**  $\text{abs}(\cos(\text{zen1})/\cos(\text{zen2})-1)$  less than 0.02; **ABI CH14 :**  $\text{std}(\text{M16}) / \text{mean}(\text{M16}) < 0.01$  ; ABI within CrIS are averaged

Spectral response functions of ABI channel 7 and 11



After gap was filled



The comparison between CrIS and ABI over the gap spectral regions also shows considerable results, **especially for channel 11**. Their differences in the channel 7 and 11 are basically similar to those observed over other infrared channels.

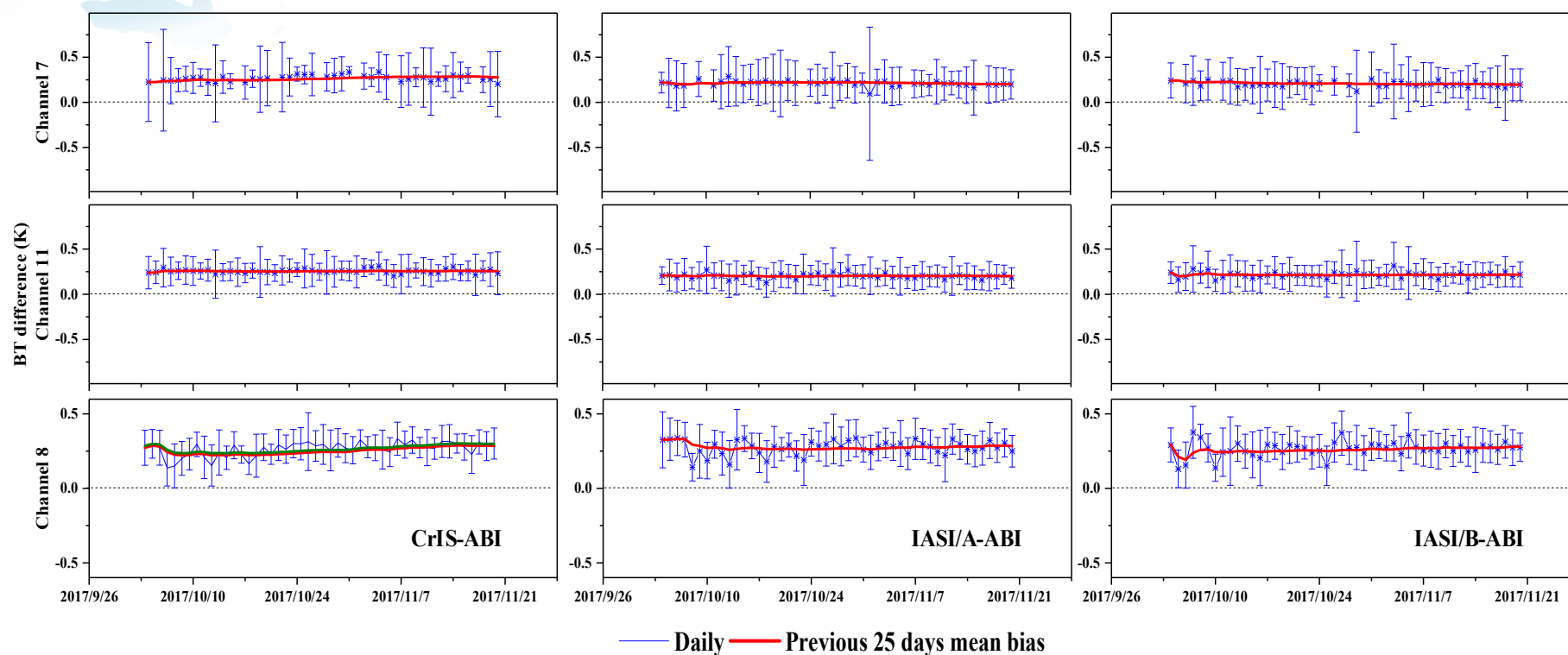


# Application



## CrIS – ABI & IASI-ABI

**SNO selection:** FOV distance : less than 7 km for CrIS & less than 6.5 for IASI; **Time difference:** less than 10 minutes ;  
**View angle difference:**  $\text{abs}(\cos(\text{zen1})/\cos(\text{zen2})-1)$  less than 0.02; **ABI CH14 :**  $\text{std}(\text{M16}) / \text{mean}(\text{M16}) < 0.01$  ; ABI within CrIS are averaged; **Nadir:** FOR 14, 15, 16 and 17 for CrIS & 13,14,15,16,17,18 for IASI



- Both the mean bias and standard deviation are in the same uncertainty level with those observed from IASI/A and B.
- This actually indicates that the prediction uncertainty is far below than the inter-comparison uncertainty when the CrIS convolved with the broadband channels.
- their mean bias also slightly decreased over the water vapor channel 8 after the little gap was filled by the proposed method, as the green line (before the gap was filled)



## Conclusions and future work



- ❖ **A new prediction algorithm was developed to fill up the CrIS gap channels** from 650 – 2755 cm<sup>-1</sup> and also can be applied to other instruments with spectral gap.
- ❖ The proposed gap channel **prediction method shows good performance** and great potential in the inter-calibration studies, especially for the “Big Gap” spectral region over VIIRS-M14 and ABI-CH11.
- ❖ The gap filling up coefficients are ready and easily for people to use and test now.
  - ❖ <ftp://ftp.orbit.nesdis.noaa.gov/pub/smcd/spb/lwang/cris-gap-coeff/>
- ❖ **Spectral gaps will be continually existed in the future hyperspectral sounders, such as, JPSS-1/2, FY-3/4?, ..., which makes the gap filling task become essentially important now.**

### Future work

- ☐ The training dataset need to be improved in next step.
- ☐ The inter-comparison performance of the predicted CrIS gap channels with other geostationary sensors, such as, AHI and the Advanced Geosynchronous Radiation Imager (AGRI) on board the recently launched FengYun-4 satellite, need to be further investigated in future studies.

# Thanks

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