

# AIRS Point Spread Function Reconstruction using AIRS and MODIS Data

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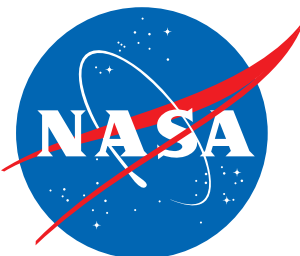
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**NASA Sounder Science Team Meeting**

**October 2 - 16, 2020**



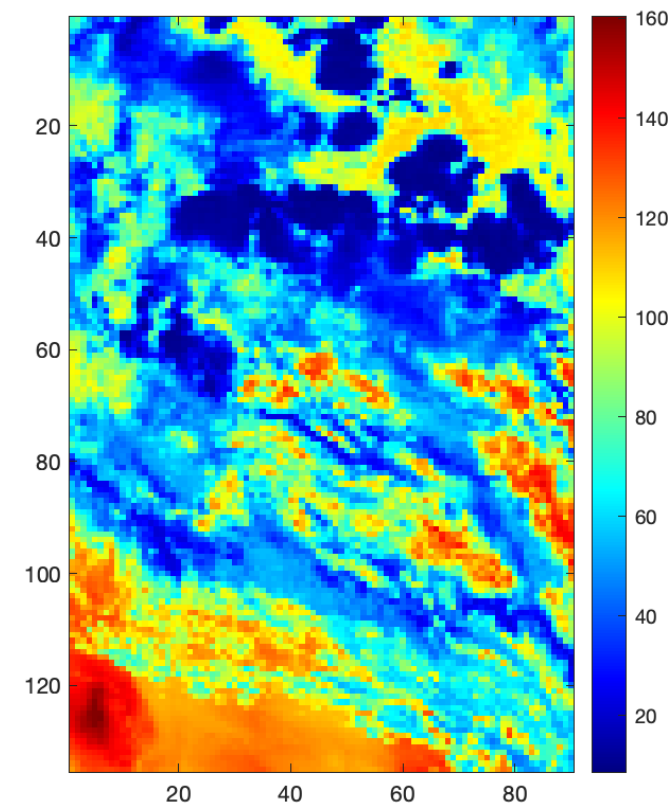
# Motivation and Approach

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- The purpose of this effort is to use MODIS data to refine our knowledge of post-launch AIRS point spread functions (PSFs), including suspected changes over the mission.
- Deriving mathematical optimization formulation for reconstruction of AIRS spatial response functions from AIRS and MODIS data.

# AIRS Data

- AIRS spatial response is different for each of 2378 channels and 90 scan angles.
- Channels range from 3.7  $\mu\text{m}$  to 15.4  $\mu\text{m}$
- Spatial resolution is 13.5 km
- Granule is 90 scan angles by 135 scans
  - 240 granules per day
- Match AIRS and MODIS radiances.
- Resample MODIS radiances onto AIRS PSF grid
  - AIRS PSFs are 39x39 pixels, 0.04 degrees / pixel
  - Each AIRS footprint corresponds to 40x40 MODIS pixels
  - MODIS channel 31 used with 1 km resolution



AIRS granule  
Channel 776 ( $913.4 \text{ cm}^{-1}$ ),  
window channel 3

# Minimization Problem

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We want to reconstruct a point spread function  $K_i$  by solving the following minimization problem:

$$\min_{K_i} E(K_i) = \min_{K_i} \left\| L'_{AIRS,i,sc}(K_i) - L'_{MODIS,i,sc} \right\|_2^2$$



# Approaches Investigated

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- Different types of regularization constraints on the reconstruction:
  - Constraint on magnitude of reconstruction
  - Constraint on  $L^2$ -norm of the gradient of reconstruction (Tikhonov regularization)

⇒ No regularization was required.
- Log barrier soft positivity constraint on the reconstruction
- Different types of optimization algorithms:
  - Regular ( $L^2$ ) Gradient Descent
  - Sobolev ( $H^1$ ) Gradient Descent
- Different initial conditions:

⇒ The methodology is not sensitive to the choice of initial conditions.

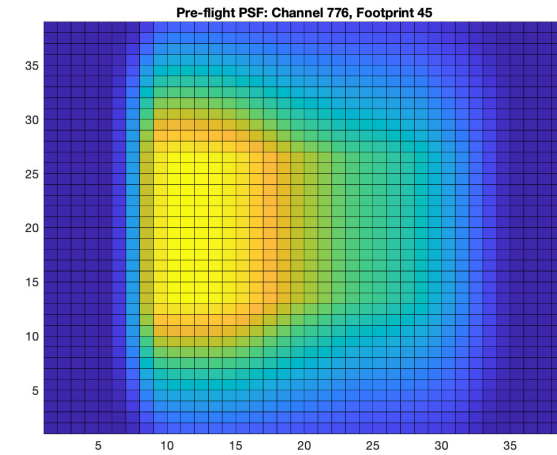
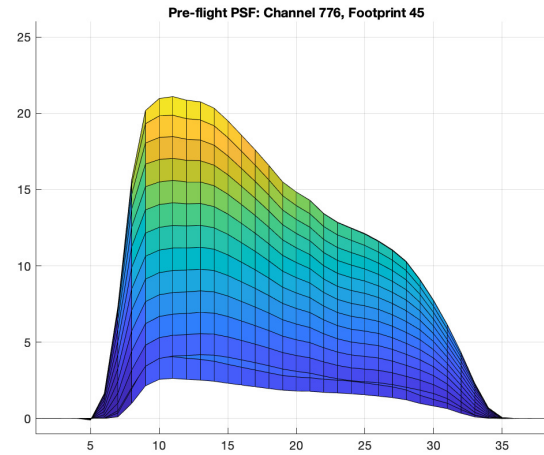
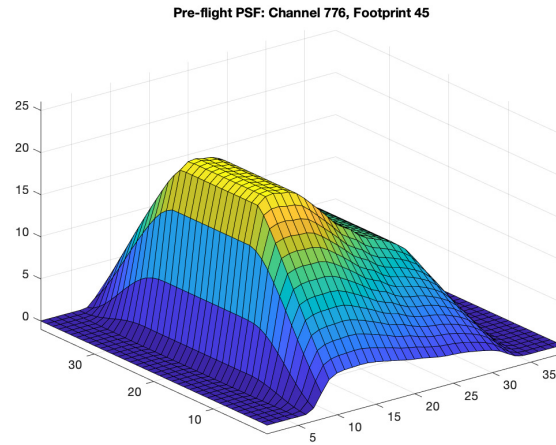
# Latest Investigations

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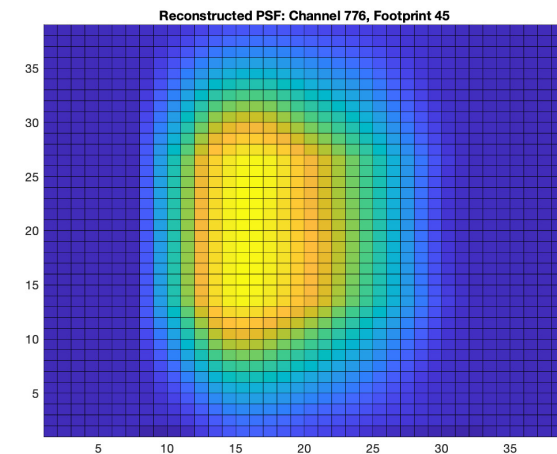
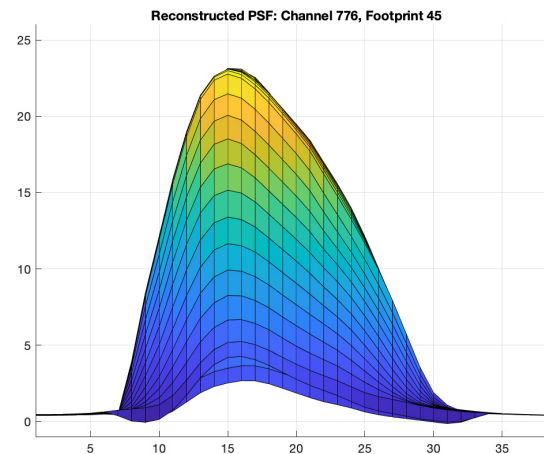
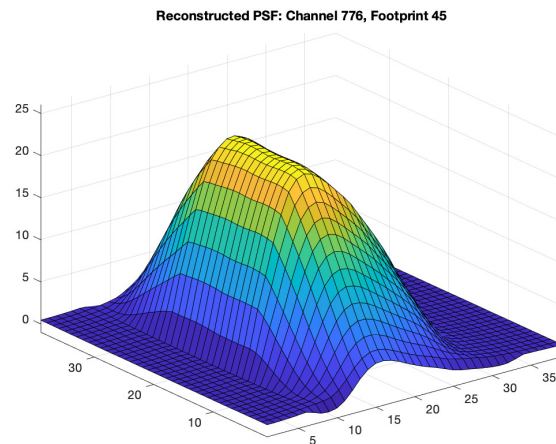
- Examined full days of data for each of the three days:
  - Data collected right after the launch:
    - March 1, 2003
    - March 2, 2003
  - Data collected at the middle of the mission:
    - March 1, 2014
- Used 235 to 239 granules for each of these days.
- Data over oceans were considered for deriving PSFs.

# Channel 776 (913.4 cm<sup>-1</sup>), Footprint 45

## Pre-flight PSF

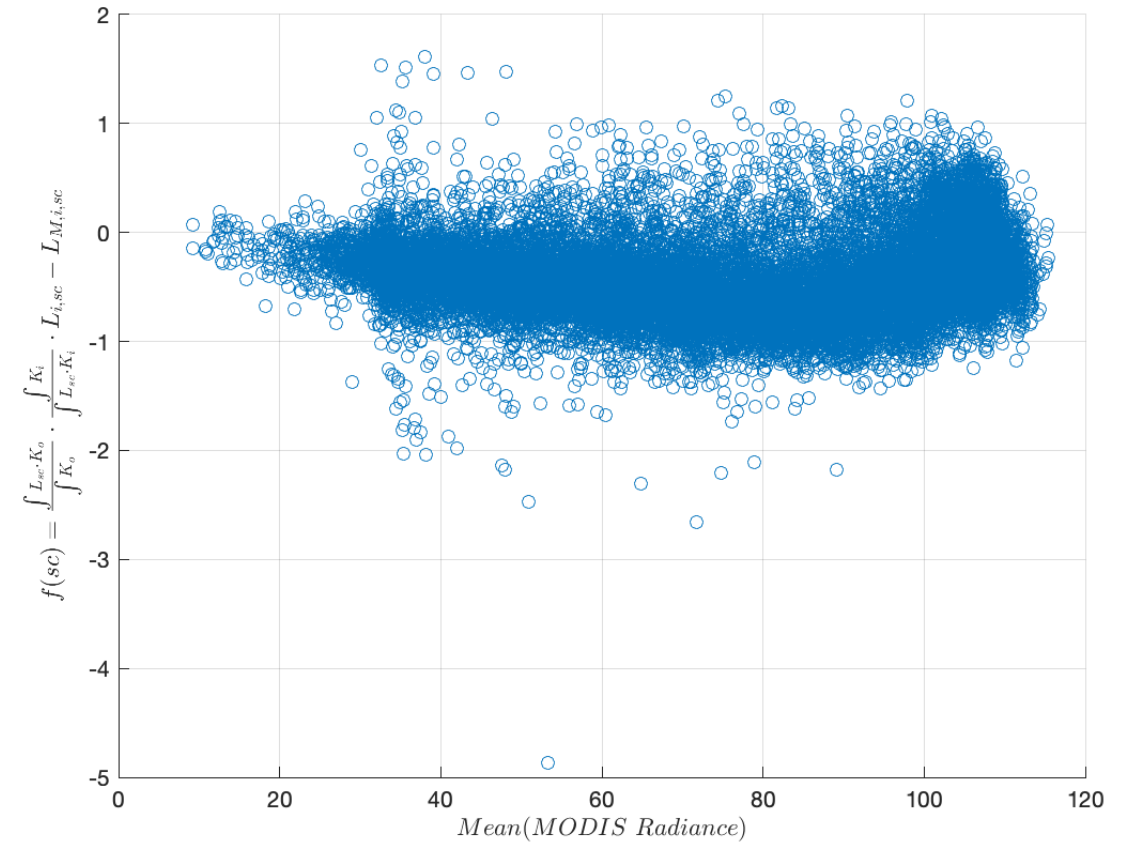
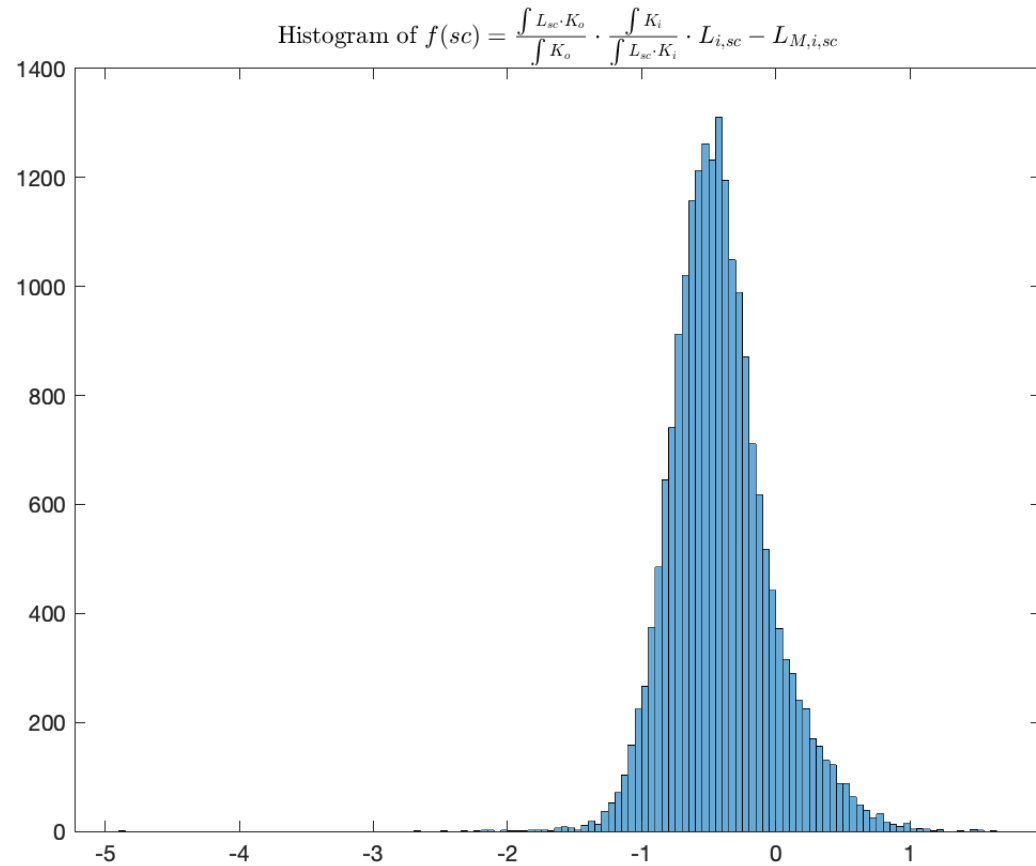


## Reconstructed PSF



Using data from March 1, 2014<sup>7</sup>

# Residuals: PSF Reconstruction for Channel 776, Footprint 45



## How does PSF trained on one day's data agree with the data for a different day?

- Repeatability: Examine data from consecutive days (March 1 & 2, 2003).
- Change: Compare data collected right after the launch (2003) with data collected at the middle of the mission (2014).

### Channel 776 (913.4 cm<sup>-1</sup>), Footprint 45

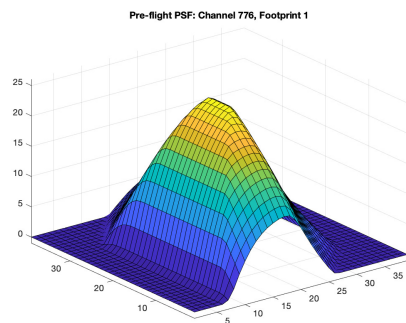
Residuals	train \ test	2003.03.01	2003.03.02	2014.03.01
	Pre-flight PSF	0.5960	0.5862	0.5251
	2003.03.01	0.3091	0.3182	0.3082
	2003.03.02	0.3092	0.3167	0.3071
	2014.03.01	0.3158	0.3237	0.3102

### Observations:

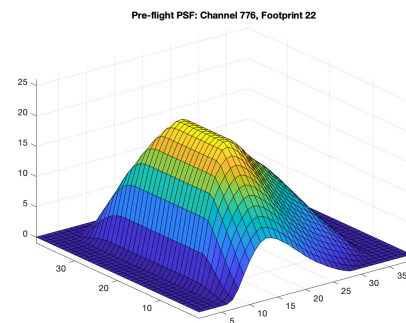
- Data from different dates agree much better with PSFs computed from any other date than it does with pre-flight PSFs.
- The 2014 data has somewhat higher residuals, suggesting some sort of degradation if we see similar effects over many channels.

# Pre-flight PSFs for channel 776 (913.4 cm<sup>-1</sup>)

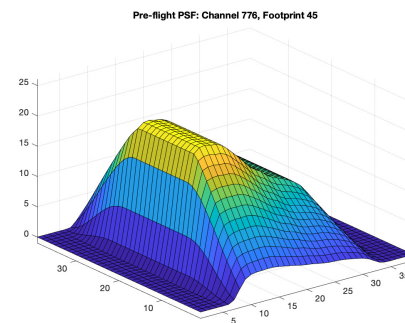
## Footprint 1



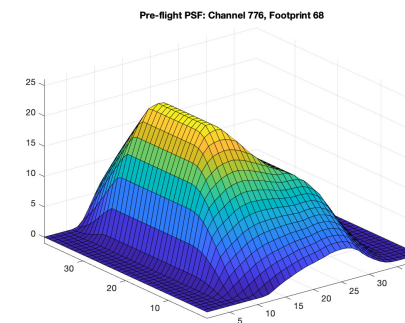
## Footprint 22



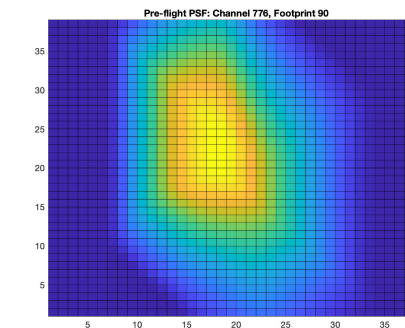
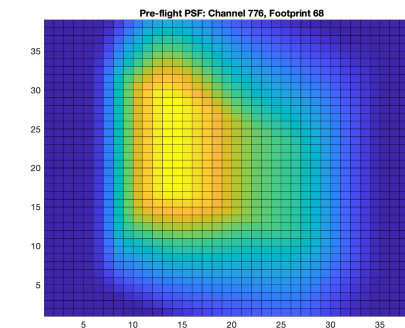
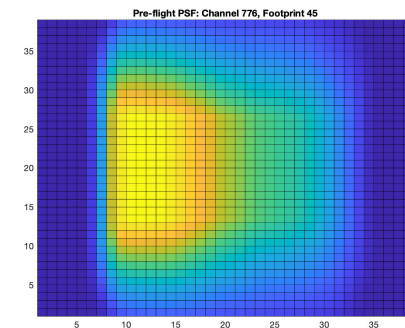
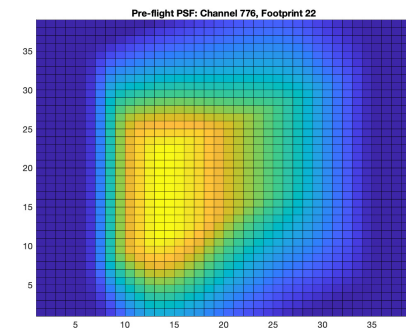
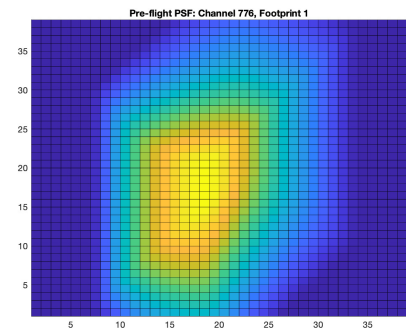
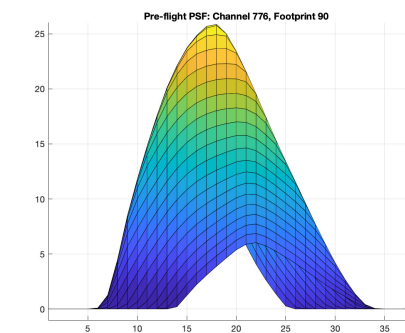
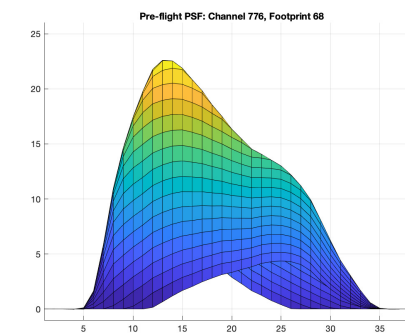
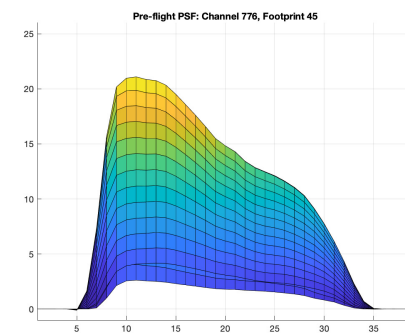
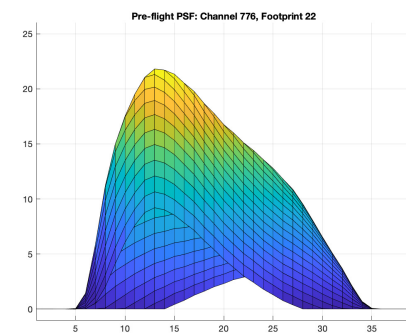
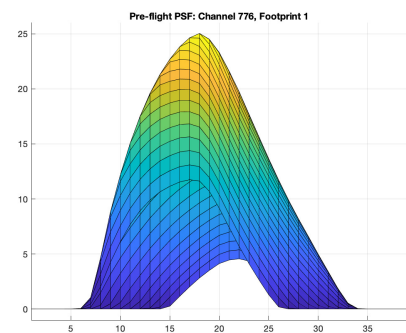
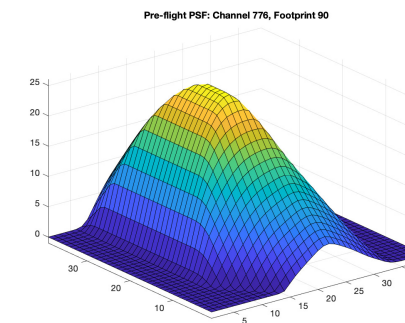
## Footprint 45



## Footprint 68



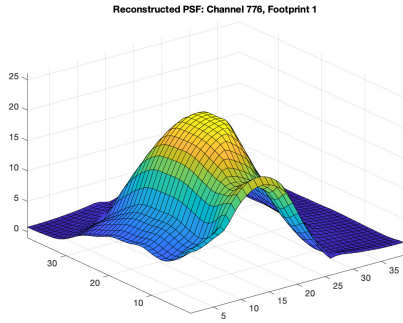
## Footprint 90



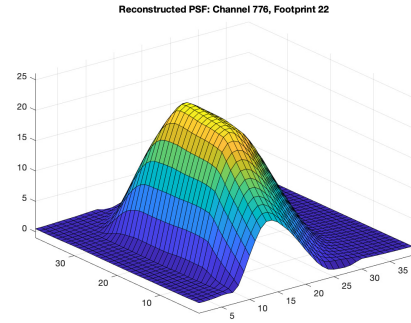


# Reconstructed PSFs for channel 776 (913.4 cm<sup>-1</sup>) based on March 1, 2014 data

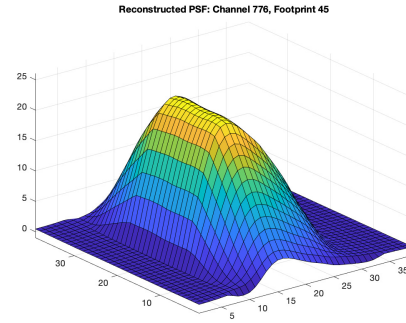
## Footprint 1



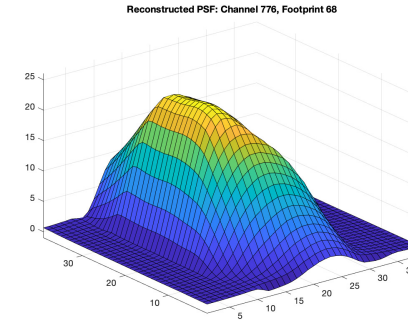
## Footprint 22



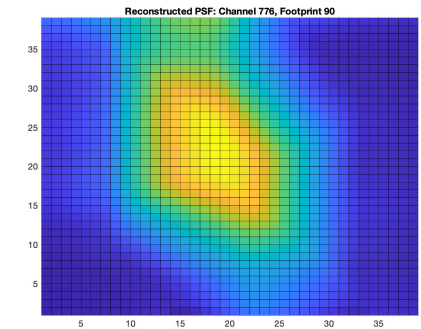
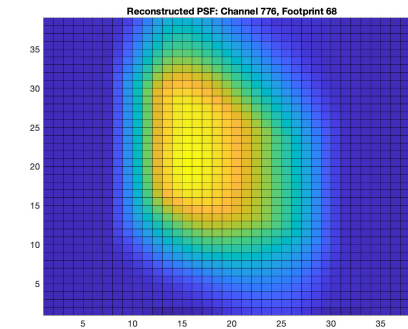
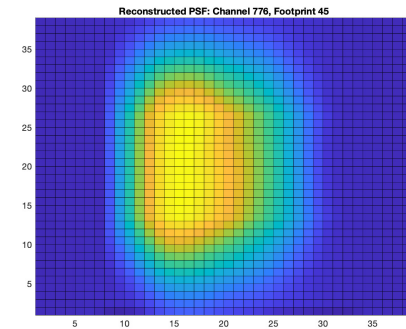
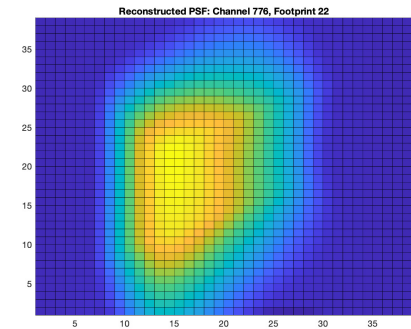
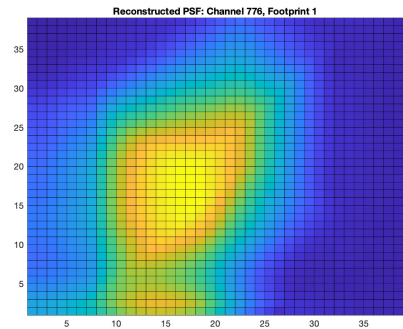
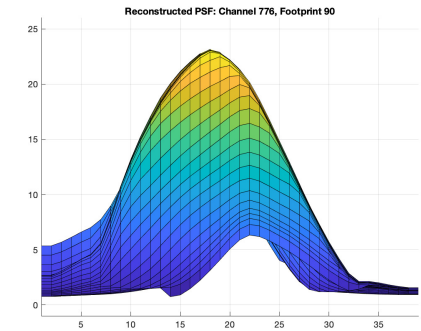
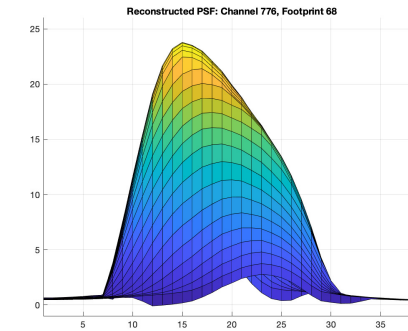
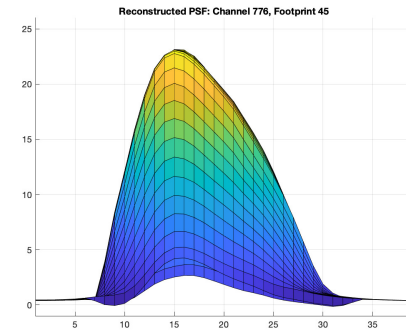
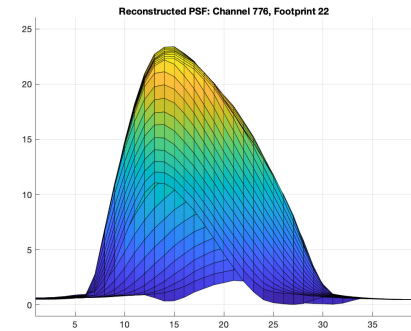
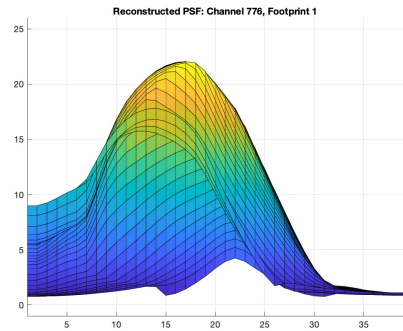
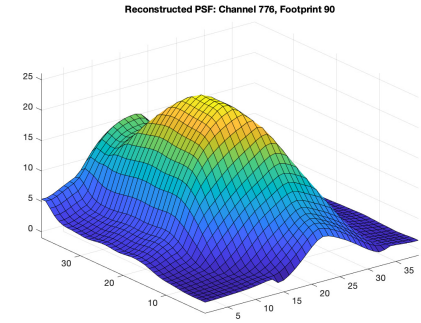
## Footprint 45



## Footprint 68



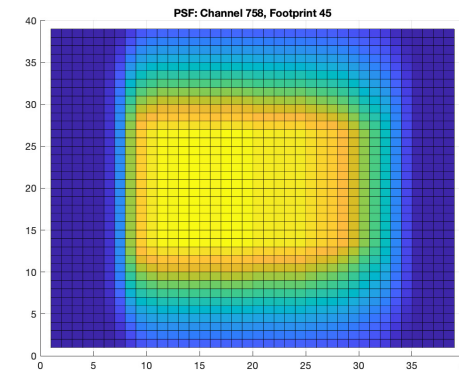
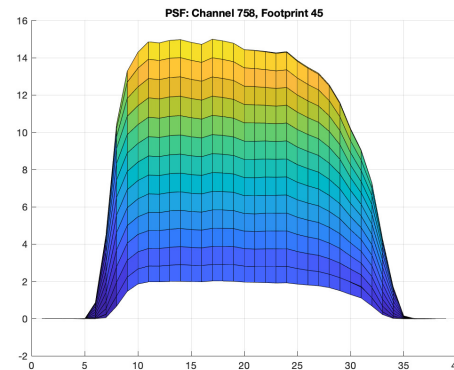
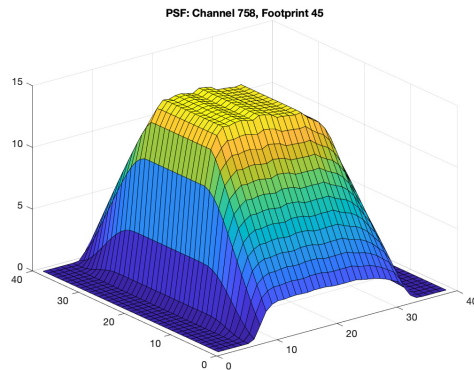
## Footprint 90



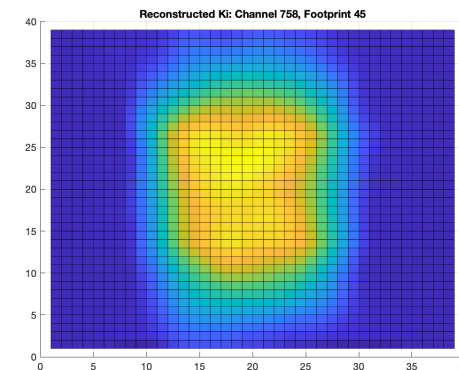
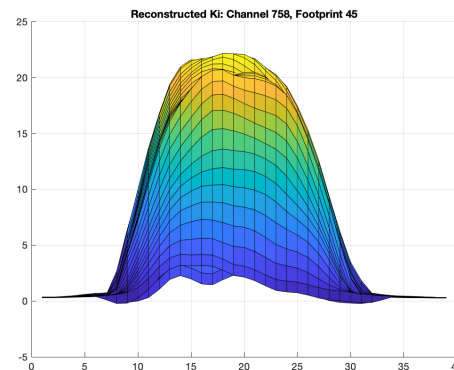
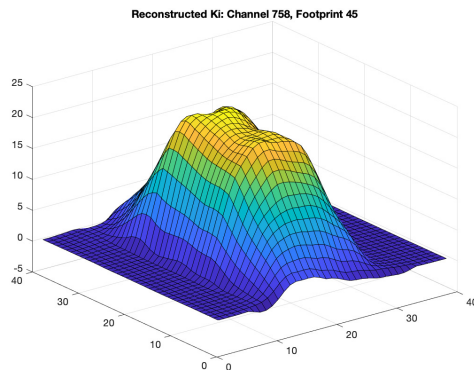
## Channel 758 (900.0 cm<sup>-1</sup>), Footprint 45

- Examine a similar window channel, now the problem module of M-08.
- As with the 913.4 cm<sup>-1</sup> channel, the reconstructed PSFs are much narrower in the X dimension.

Pre-flight PSF



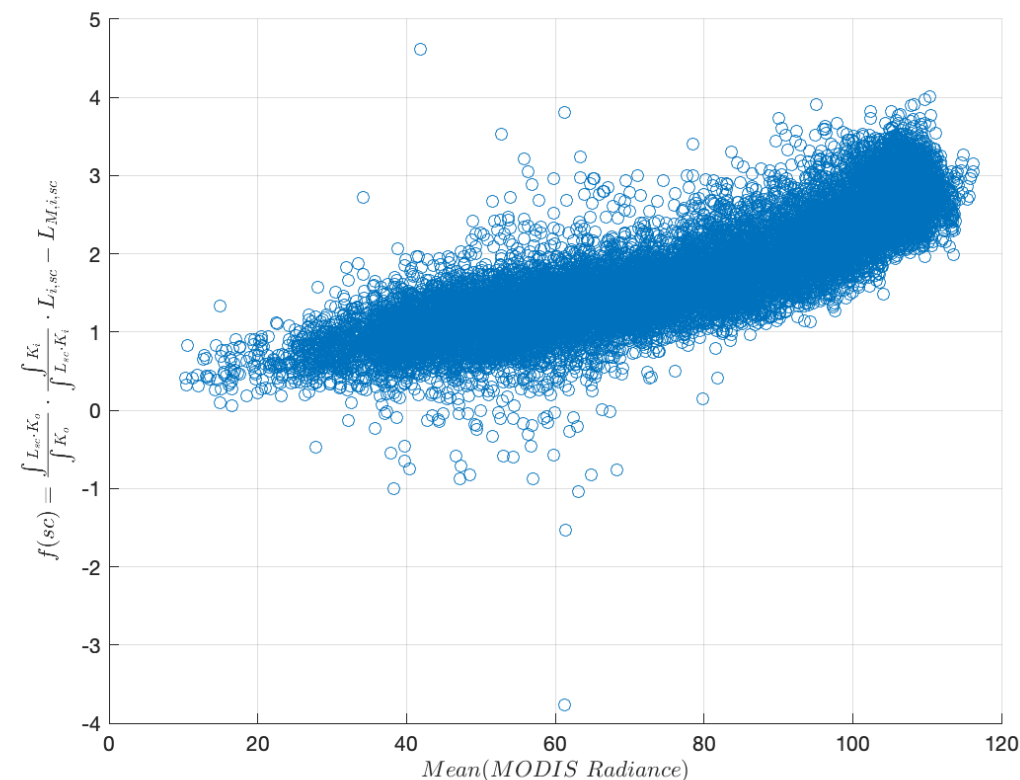
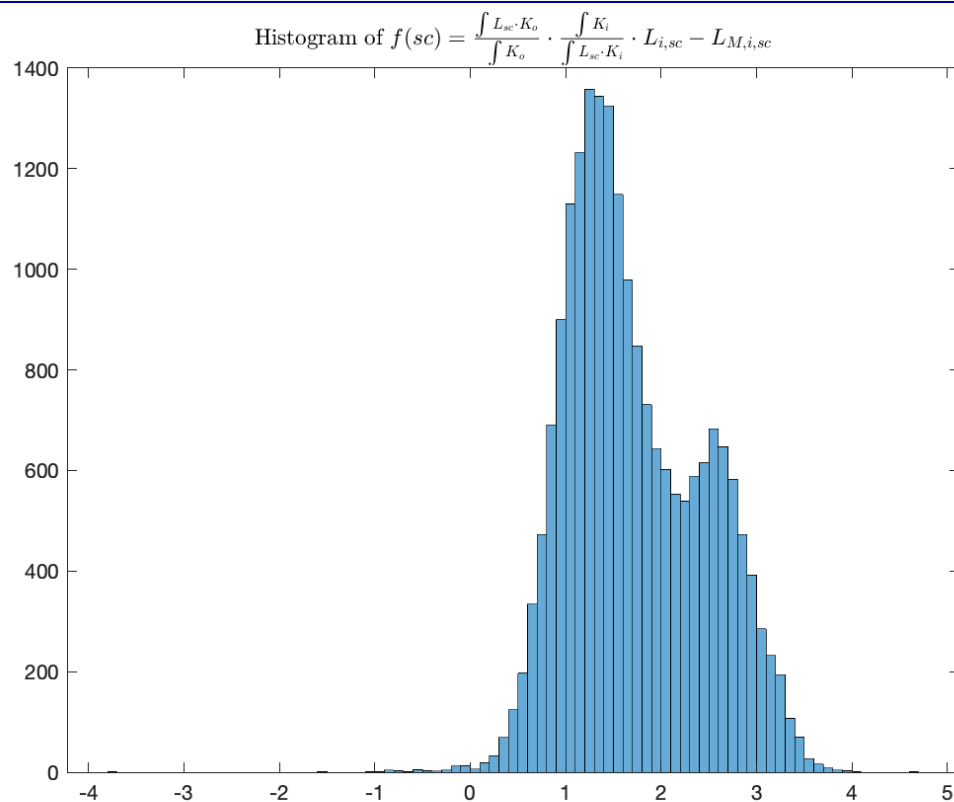
Reconstructed PSF



Using data from  
March 1, 2014



# Residuals: PSF Reconstruction for Channel 758, Footprint 45



- M-08 is already known to have a bias with scene temperature, and now we see it when we compare it to MODIS.
- Residuals are much larger for the M-08 channel, especially at higher scene brightness. This might suggest some stray signal that can't be accounted for by 40x40 MODIS pixels – perhaps out-of-band or out-of-area.

## How does PSF trained on one day's data agree with the data for a different day?

### Channel 758 (900.0 cm<sup>-1</sup>), Footprint 45

Residuals	train \ test	2003.03.01	2003.03.02	2014.03.01
	Pre-flight PSF	3.7407	3.5915	3.7845
	2003.03.01	3.4131	3.2919	3.4909
	2003.03.02	3.4178	3.2831	3.4861
	2014.03.01	3.4368	3.3067	3.4674

### Observations:

- Residuals are much higher for 900.0 cm<sup>-1</sup> compared to 913.4 cm<sup>-1</sup> channel, but we still improve over pre-flight PSF.
- 2014 is again worse than 2003, supporting the idea of degradation.

# Discussions of Results

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- Approach is successful:
  - Generates PSF with smaller residuals compared to pre-flight PSF.
- Soft positivity constraint on reconstruction:
  - Does not prevent negative values on edges of PSF (similar to pre-flight PSF).
- Sidelobes at ends of scan (footprints 1 and 90) – future investigation.
- The narrowing we saw at nadir and the nodes at the ends of the scan. Re-check the MODIS indexing.
- M-08 A/B currently does not perform as well as channel 776.
- The new PSFs may have time dependence.
- The new reconstructed PSFs should be a big help to analyses using AIRS and MODIS data together.

# Future Work

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- Reconstruct and examine PSFs for different channels.
  - Starting to investigate M-08 issues.
  - Look at shortwave channels. (Shortwave trends in the radiances).
- Publish a better set of reconstructed PSFs.

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# Backup Slides

# AIRS Radiances

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The spatially averaged radiance from AIRS depends on the scene and AIRS spatial response:

$$L_{AIRS,i,sc} = \frac{\sum_{x,y} L_{i,sc}(x,y) K_i(x,y)}{\sum_{x,y} K_i(x,y)},$$

where

$L_{i,sc}$  = scene radiance in channel  $i$  at scan  $sc$

$K_i$  = AIRS spatial response function for a given footprint (scan angle)

$L_{AIRS,i,sc}$  = AIRS L1B radiance in channel  $i$  at scan  $sc$

# AIRS Radiances

The spatially averaged radiance from AIRS depends on the scene and AIRS spatial response:

$$L_{AIRS,i,sc} = \frac{\sum_{x,y} L_{i,sc}(x,y) K_i(x,y)}{\sum_{x,y} K_i(x,y)},$$

Given radiance  $L_{AIRS,i,sc}$  captured by AIRS instrument, we can correct it, obtaining  $L'_{AIRS,i,sc}$ :

$$L'_{AIRS,i,sc} = \frac{\sum_{x,y} L_{sc}(x,y) K_o(x,y)}{\sum_{x,y} K_o(x,y)} \frac{\sum_{x,y} K_i(x,y)}{\sum_{x,y} L_{sc}(x,y) K_i(x,y)} L_{AIRS,i,sc},$$

where

$L_{sc}$  = MODIS scene radiance at scan  $sc$

$K_o$  = average AIRS spatial response function (over all channels)

$L'_{AIRS,i,sc}$  = spatially corrected AIRS radiance in channel  $i$  at scan  $sc$

# MODIS Radiances

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The MODIS averaged radiance  $L'_{MODIS,i,sc}$  (to compare with  $L'_{AIRS,i,sc}$ ) must also be weighted by the average AIRS spatial response function:

$$L'_{MODIS,i,sc} = \frac{\sum_{x,y} L_{sc}(x,y) K_o(x,y)}{\sum_{x,y} K_o(x,y)}.$$



# Minimization Problem

---

We want to find  $K_i$  by solving the following minimization problem:

$$\min_{K_i} E(K_i) = \min_{K_i} \left\| L'_{AIRS,i,sc}(K_i) - L'_{MODIS,i,sc} \right\|_2^2$$

We know that the point spread function  $K_i$  should be non-negative.

⇒ Introduce **log barrier** soft positivity constraint on  $K_i$ . The new minimization problem is:

$$\min_{K_i} E(K_i) = \min_{K_i} \left\| L'_{AIRS,i,sc}(K_i) - L'_{MODIS,i,sc} \right\|_2^2 - \lambda \int \log(K_i) dx$$

# Minimization Problem

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or

$$\min_{K_i} E(K_i) = \min_{K_i} \sum_{sc} \left( \frac{\int L_{sc}(x) K_o(x)}{\int K_o(x)} \cdot \frac{\int K_i dx}{\int L_{sc} \cdot K_i dx} \cdot L_{AIRS,i,sc} - L'_{MODIS,i,sc} \right)^2 - \lambda \int \log(K_i) dx.$$

# Euler-Lagrange Equation

Minimization problem:

$$\min_{K_i} E(K_i) = \min_{K_i} \sum_{sc} \left( \frac{\int L_{sc}(x) K_o(x)}{\int K_o(x)} \cdot \frac{\int K_i dx}{\int L_{sc} \cdot K_i dx} \cdot L_{AIRS,i,sc} - L'_{MODIS,i,sc} \right)^2 - \lambda \int \log(K_i) dx.$$

$\Rightarrow K_i$  must solve the following Euler-Lagrange equation:

$$\begin{aligned} \partial_{L^2} E(K_i) &= \sum_{sc} \left( \left[ \frac{\int L_{sc}(x) K_o(x)}{\int K_o(x)} \cdot \frac{\int K_i}{\int L_{sc} K_i} \cdot L_{AIRS,i,sc} - L'_{MODIS,i,sc} \right] \right. \\ &\quad \times \left. \frac{\int L_{sc}(x) K_o(x)}{\int K_o(x)} \cdot L_{AIRS,i,sc} \cdot \frac{\int L_{sc} K_i - (\int K_i) L_{sc}}{(\int L_{sc} \cdot K_i)^2} \right) - \lambda \frac{1}{K_i} = 0. \end{aligned}$$

# Sobolev Gradient Descent

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Advance the Euler-Lagrange equation using Sobolev Gradient Descent:

$$\frac{dK_i}{dt} = -\partial_{H^1} E(K_i) = -(I - \Delta)^{-1} \partial_{L^2} E(K_i),$$

which can be re-written as:

$$(I - \Delta) \frac{dK_i}{dt} = -\partial_{L^2} E(K_i)$$

# Acknowledgements

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**Thank You!**