



# MULTI-SENSOR DATA PROCESSING METHOD FOR IMPROVED SATELLITE RETRIEVALS

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## 1. INTRODUCTION

A growing number of satellite sensors provide large amounts of data in reflective solar bands. These data improve the temporal coverage of earth observations. It is now a high priority to develop robust methods to maximize the use of these archived data for satellite retrievals. The author primarily focuses on the following issues using multi-sensor data.

- sensor calibration <satellite level>
- aerosol optical depth <atmosphere level>
- NDVI consistency <surface level>

## 2. METHODS

Radiative transfer equation can be written as

$$\rho_{TOA} = T_g \left( \rho_a + \frac{\rho T_s T_v}{1 - \rho S} \right) \quad (1)$$

where  $\rho_{TOA}$ —apparent reflectance,  $\rho_a$ —path reflectance,  $\rho$ —surface reflectance,  $T_g$ —atmospheric absorption,  $T_s$ — $T_v$ —atmospheric scattering,  $S$ —atmospheric albedo.

Parameters values in EQ.(1) differ with sensor type, atmospheric and surface conditions. The  $\rho_a$ ~ $T_g$ ~ $T_s$ ~ $T_v$ ~ $S$  are functions of atmospheric parameters (including AOT). The  $\rho_{TOA}$  depends on sensor calibration. If differences in dual-sensor  $\rho$  data are eliminated, EQ.(2) holds.

$$\rho_{TOA,2} = \underbrace{\left( f_{2/1} \times \frac{T_{g2} T_{s2} T_{v2}}{T_{g1} T_{s1} T_{v1}} \right)}_a \times \rho_{TOA,1} + T_{g2} \left( \underbrace{\rho_{a,2} - f_{2/1} \times \rho_{a,1} \times \frac{T_{s2} T_{v2}}{T_{s1} T_{v1}}}_b \right)$$

where **f** accounts for BRDF and sensor differences in surface reflectances, a and b are the two coefficients. (2)

EQ.(2) is the relationship between dual-sensor observations, then the NDVI relationship is

$$NDVI_2 = \frac{(a_N + a_R) NDVI_1 + (a_N - a_R) + 2(b_N - b_R) / (R_1 + NIR_1)}{(a_N + a_R) + (a_N - a_R) NDVI_1 + 2(b_N + b_R) / (R_1 + NIR_1)} \quad (3)$$

Simplifying (3) yields  
where n are coefficients.

$$NDVI_2 = \frac{n_1 \times NDVI_1 + n_2}{n_1 + n_2 \times NDVI_1} \quad (4)$$

## 3. RESULTS

**3.1 Sensor intercalibration:** know atmospheric and surface conditions—solve for sensor parameters

Study area: Dunhuang calibration site, China  
Sensors: TERRA MODIS and NOAA-17 AVHRR

The parameter f calculated with desert reflectance

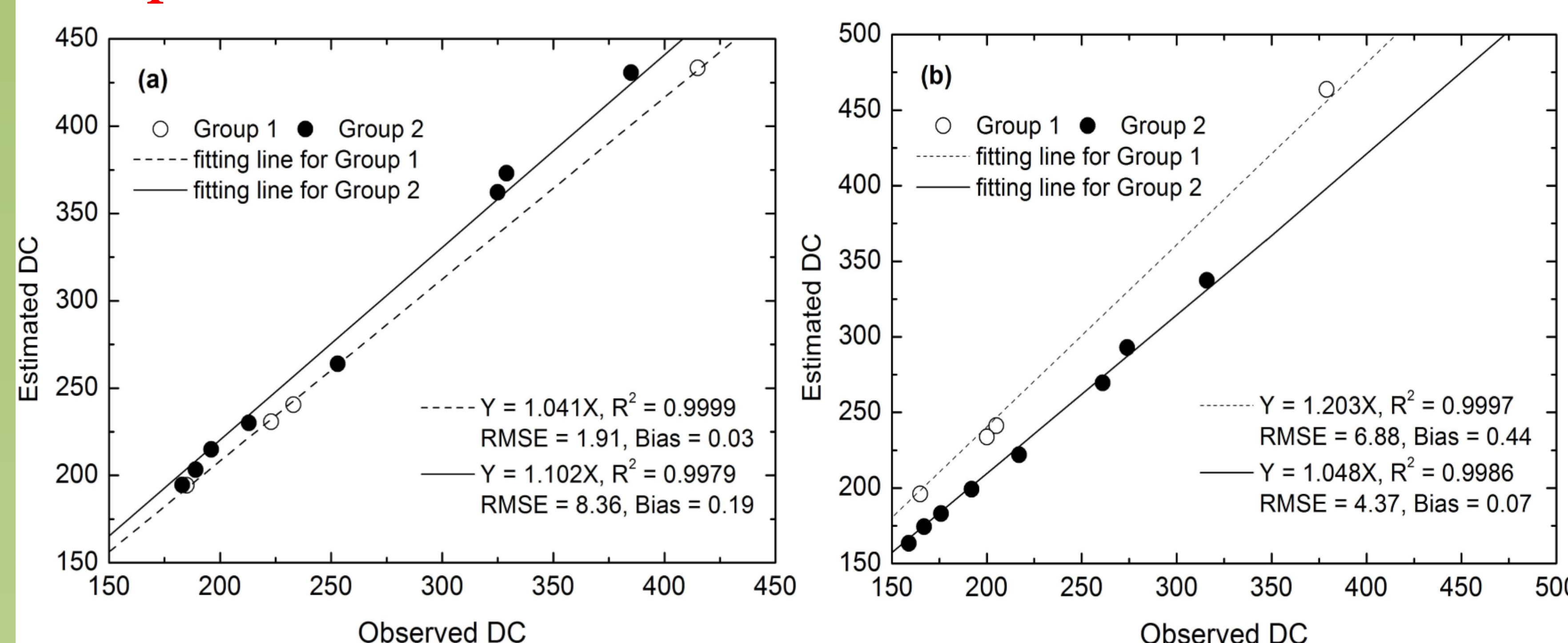


Figure 1. Calibration results

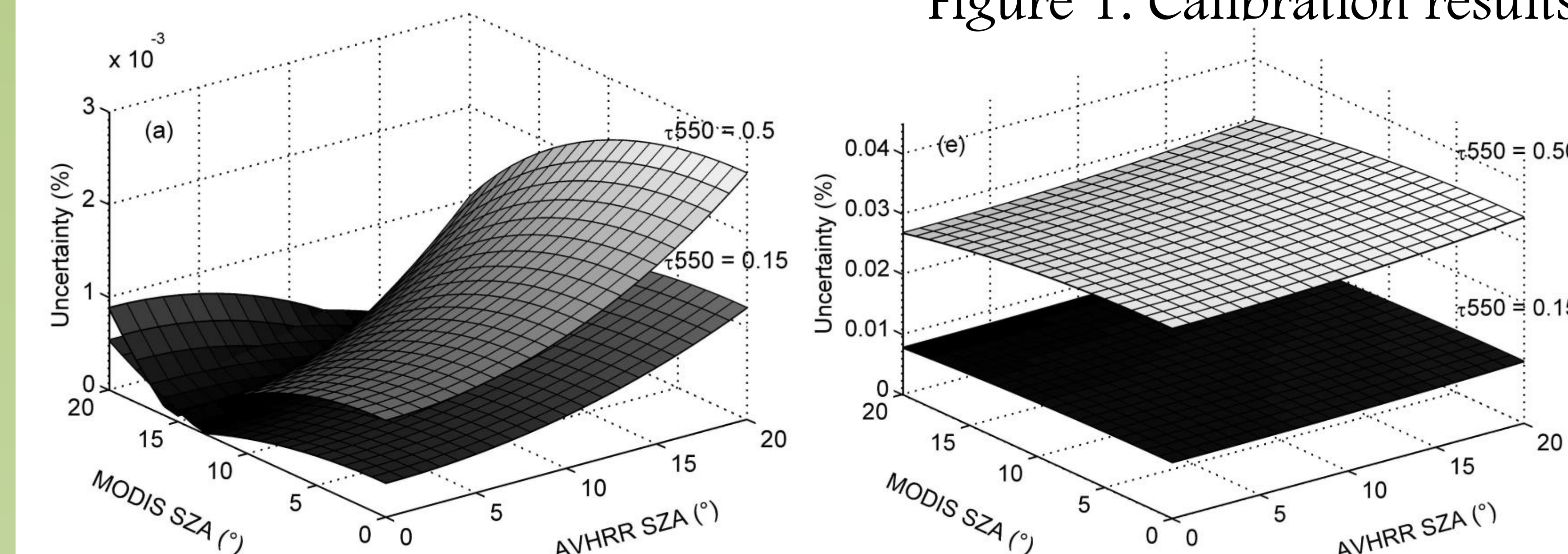


Figure 2. The effects of AOT

**3.2 Aerosol retrieval:** know sensor and surface conditions—solve for atmospheric parameters

Study area: Poyang Lake basin, China  
Sensors: TERRA MODIS and AQUA MODIS

The parameter f calculated as SWIR band ratio

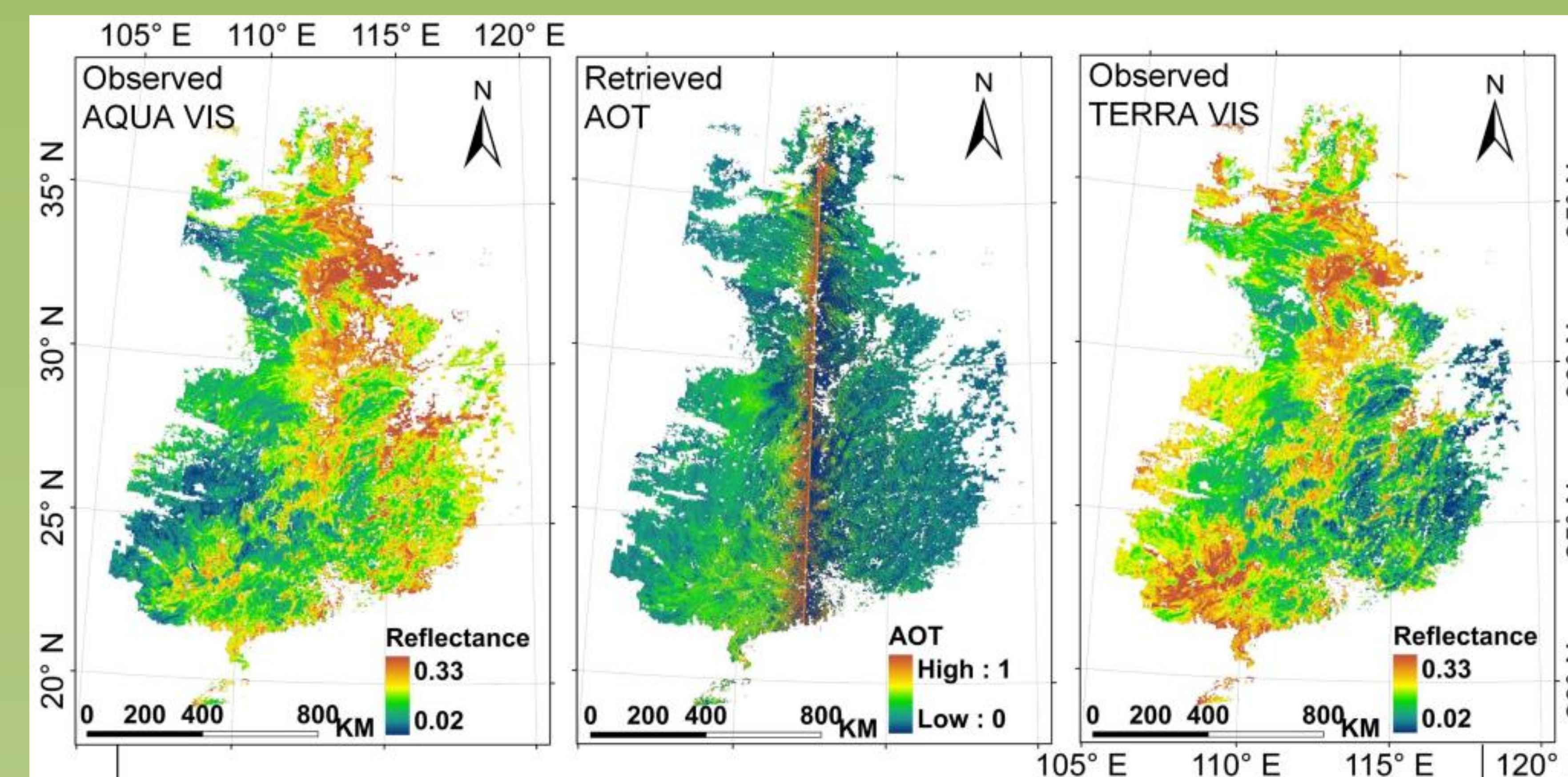


Figure 3. Comparison of aerosol data from (a) TERRA, (b) this method and (c) AQUA.

**3.3 NDVI consistency:** know sensor and atmospheric conditions—solve for surface parameters

Study area: the whole globe

Sensors: MODIS/NOAA-x AVHRR/NPP VIIRS

The parameter f calculated from MODIS reflectance

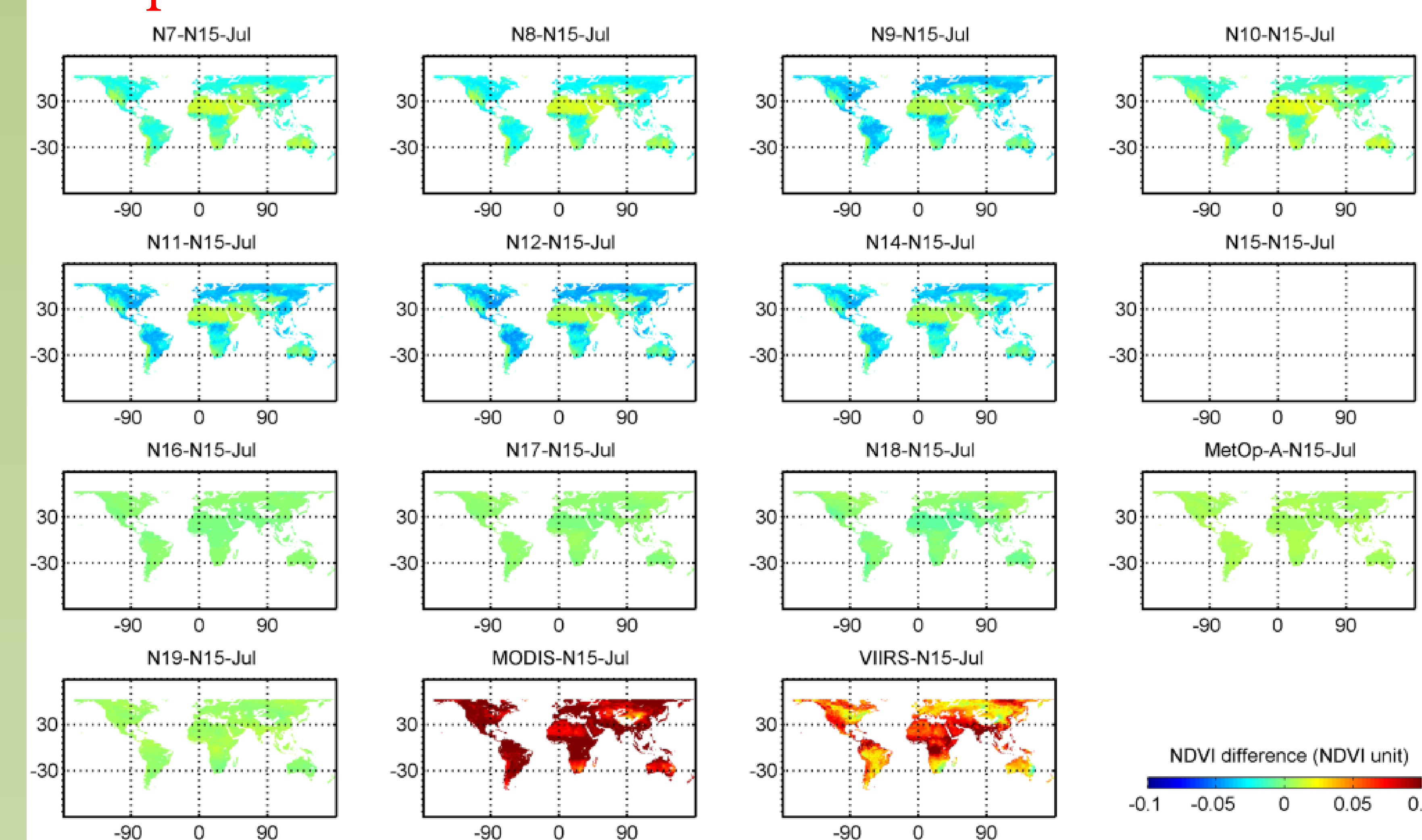


Figure 4. Simulated NDVI differences in relative to MODIS (July)

## 4. DISCUSSION AND CONCLUSIONS

- The RT methods quantified the effects of sensor, atmospheric and surface parameters on dual-sensor data.
- At site scale, we derived quality assured sensor calibration coefficients and quantified the total uncertainty.
- At basin scale, we obtained finer-resolution and more accurate aerosol retrievals than the MODIS product.
- At global scale, we analyzed the patterns of NDVI differences among moderate-resolution satellite sensors.