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Algorithm Generalized Split-Window

MSG/SEVIRI & AVHRR/Metop

Two-Channel → 10.7 μm & 12.0 μm

$$LST = (A_1 + A_2 \frac{1-\epsilon}{\epsilon} + A_3 \frac{\Delta\epsilon}{\epsilon^2}) \frac{T_{10.7} + T_{12.0}}{2} + (B_1 + B_2 \frac{1-\epsilon}{\epsilon} + B_3 \frac{\Delta\epsilon}{\epsilon^2}) \frac{T_{10.7} - T_{12.0}}{2} + C$$

Parameters A_k , B_k & C depend on:

- Total Column Water Vapour**
- From Numerical Weather Prediction Models (ECMWF)
- Satellite View Angle**

(Generalized Split Window developed for MODIS and adapted to SEVIRI-MSG – Freitas et al., 2010)

LST Uncertainty

$$S_{LST}^2 = \sum_i \left(\frac{\partial f}{\partial X_i} \right)^2 \sigma_{X_i}^2 + \sum_j \left(\frac{\partial f}{\partial \theta_j} \right)^2 \sigma_{\theta_j}^2 + \Delta LST^2$$

Algorithm uncertainty [depend on retrieval conditions → total optical path]

Errors in algorithm parameters [depend on implicit input variables → column water vapour; view angle; land cover]

Errors in explicit algorithm inputs [sensor noise; emissivity]

SEVIRI 15 min Product

Land Surface Temperature

LST 14 Oct 2017 11:00

Uncertainty

LST Uncertainty 14 Oct 2017 11:00

SEVIRI/MSG

- Nadir pixel sampling distance: 3km
- 15 min; Available since 2005
- Re-processing in 2015: 2004 – 2012 using current LST Algorithm
- 10-daily maximum / median Available since 2012 (internal)
- 30-daily maximum / median Available since 2012 (internal)

Largest uncertainty in LST retrievals associated to land surface emissivity. Impact is largest under dry conditions

Other sources of uncertainty not quantified in the LST Error Bar

- Cloud clearing
- Aerosol load
- Non-isotropy of LST

Validation

- Comparison against in situ (reference) data
- Comparison with similar products from other sensors – for consistency assessment purposes and complementary to ground data.

In Situ Measurements Evora, Southern Portugal: Oak Trees

Radiometric temperature (°C) at Évora in a summer day: sunlit ground (red dots); tree canopy (green dots); shaded ground (black dots). The near Surface air temperature is also shown (°C; blue dots).

Upscaling

Idealized single tree view at Évora: Nadir & SEVIRI view at different local times in July.

Geometric Model – To estimate shapes of objects seen by the sensor

Boolean model – To derive overlap probabilities and the actual fraction of each end-member

Validation – Upscaling issues are more easily solved in areas with more homogeneous landscapes; see LST validation with desert station in Gobabeb (Goettsche et al., 2016).

October 2011 – September 2012

SEVIRI LST (y-axis) versus ground estimates (x-axis) obtained using the geometric model of Evora site and measurements of sunlit/shaded ground and tree canopy.

SEVIRI vs In situ Bias / Standard Deviation of differences (°C)

| | Daytime | Night-time |
|---|----------|------------|
| Upscaling: Simple average of in situ measurements | -1.2/2.2 | -0.1/1.2 |
| Upscaling: Geometric Model | 0.5/1.4 | 0.1/1.2 |

Modelling Directional Effects in LST

- LST variability with view angle depends on local spatial heterogeneity.
- Different models (e.g. Vinikov et al, 2012) calibrated with MODIS & geostationary (MSG, GOES-E and MTSAT) LST fields collocated in space and time

Model proposed by Vinikov et al. (2012)

$$\frac{T(\theta_v, \theta_i, \Delta\phi)}{T_0} = 1 + A\Phi(\theta_v) + D\Psi(\theta_v, \theta_i, \Delta\phi)$$

Solar Kernel: $\Phi(\theta_v) = 1 - \cos(\theta_v)$

Emissivity Kernel: $\Psi(\theta_v, \theta_i, \Delta\phi) = \sin(\theta_v) \cos(\theta_i) \sin(\theta_i) \cos(\Delta\phi) \cos(\theta_i - \theta_v)$

$T(\theta_v, \theta_i, \Delta\phi)$: LST at any illumination & view angles
 T_0 : LST measured from nadir view

Local Time: ~16:00 ~00:00 ~08:00

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Merged LST Product: GOES-E + SEVIRI/MSG + MTSAT

Estimated difference to "nadir-view" LST provided by the Kernel model.

- Most users consider LST regardless of its variability with view angle: Directional Effects correspond to a source of uncertainty in those conditions.
- Directional effects in LST can be represented by parametric models: Kernel (Vnikov et al., 2012); Other models are being tested.
- The goal is to estimate the uncertainty of LST associated to directional effects, i.e., the expected deviation to nadir view LST.
- At continental (global) scale, model calibration is sensitive to inconsistencies among the collocated LST data due to: differences in algorithms, input data, observation time.

References

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