



Uncertainty in evapotranspiration mapping from thermal infrared remote sensing data with EVASPA

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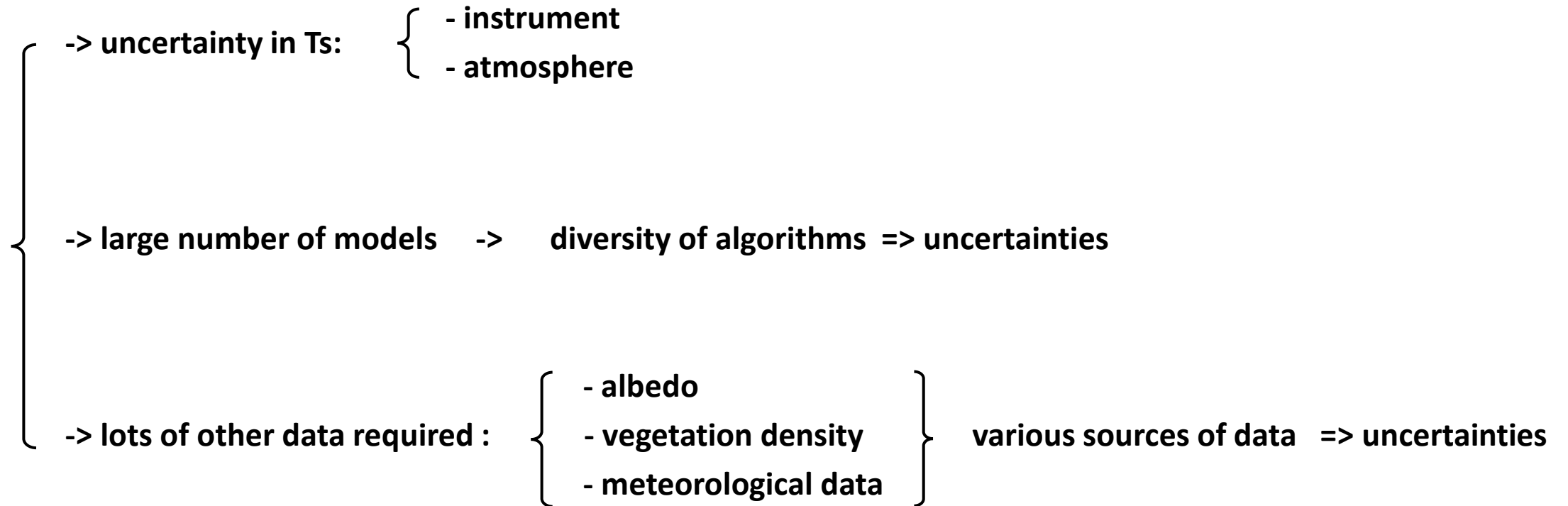
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Evapotranspiration (ET) can be derived using various models based on **thermal infrared data**



Many unknowns remain concerning the uncertainties in the derivation of ET, in particular for discriminating uncertainties from input data and models

Models of Latent Heat Flux (LE)

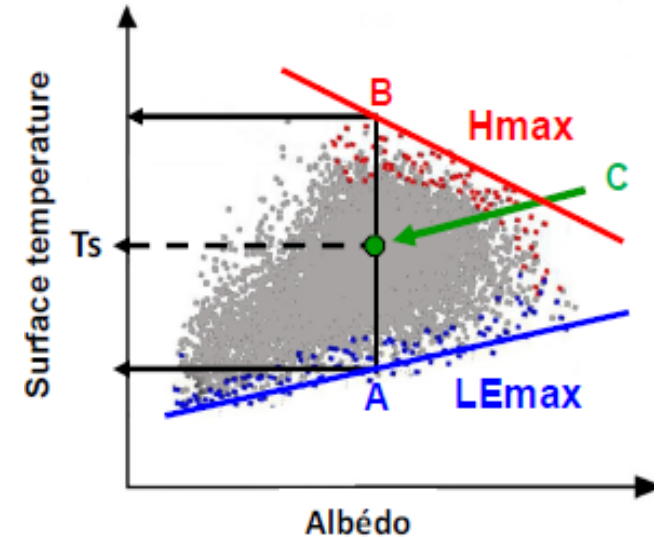
[see Lagouarde and Boulet 2016]

-> Contextual model : $LE \sim EF \times (Rn - G)$

EF = evaporative fraction <- T_s vs. *albedo or NDVI*

Rn = net radiation <- { albedo, emissivity, T_s ,
solar irradiance
atmospheric irradiance

G = ground heat flux <- Rn , $NDVI$, $fCOVER$



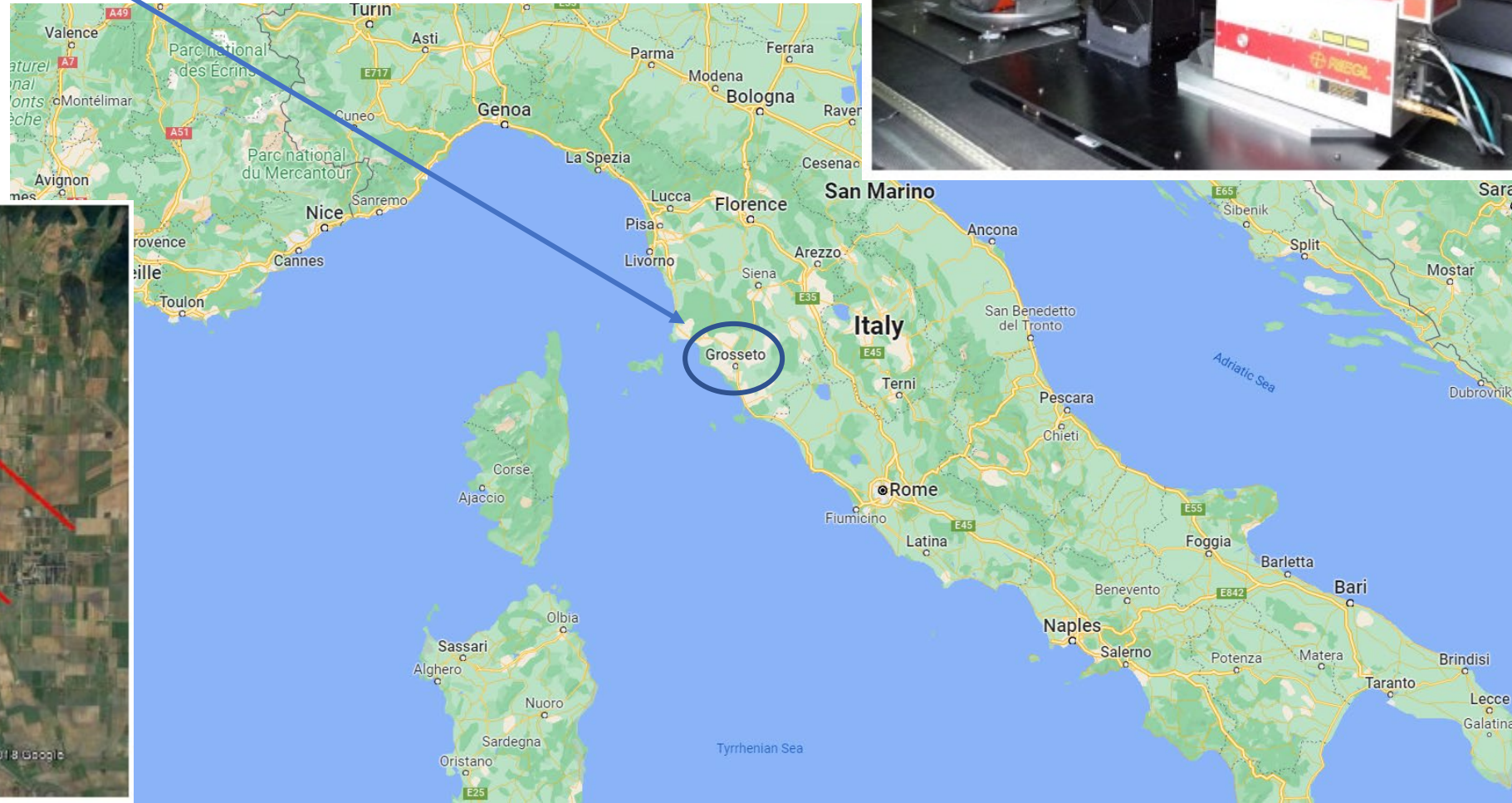
ex: S-SEBI (Roerink et al. 2000)

ESA experiment in Grosseto (Italy) in support of the LSTM program : July 2018

Airborne acquisition with TASI (TIR) and HyPLANT (VNIR, SWIR) sensors
3.5 m resolution at the ground



Agricultural area (~10 X 10 km)



Data: Grosseto experiment, July 2018

Surface temperature and emissivity : {
-> TASI data provided by ESA
-> TASI data re-processed at Univ. of Valencia

Surface spectral reflectances : -> HyPlant data {
-> NDVI
-> albedo
-> LAI
-> roughness

Incident radiations {
-> ERA 5 reanalysis (ECMWF – COPERNICUS Climate change Service)
-> SoDa Meteosat (COPERNICUS Atmosphere Monitoring Service)
-> COSMO6-REA Reanalysis (Bonn University)
-> in situ data (2 instruments)

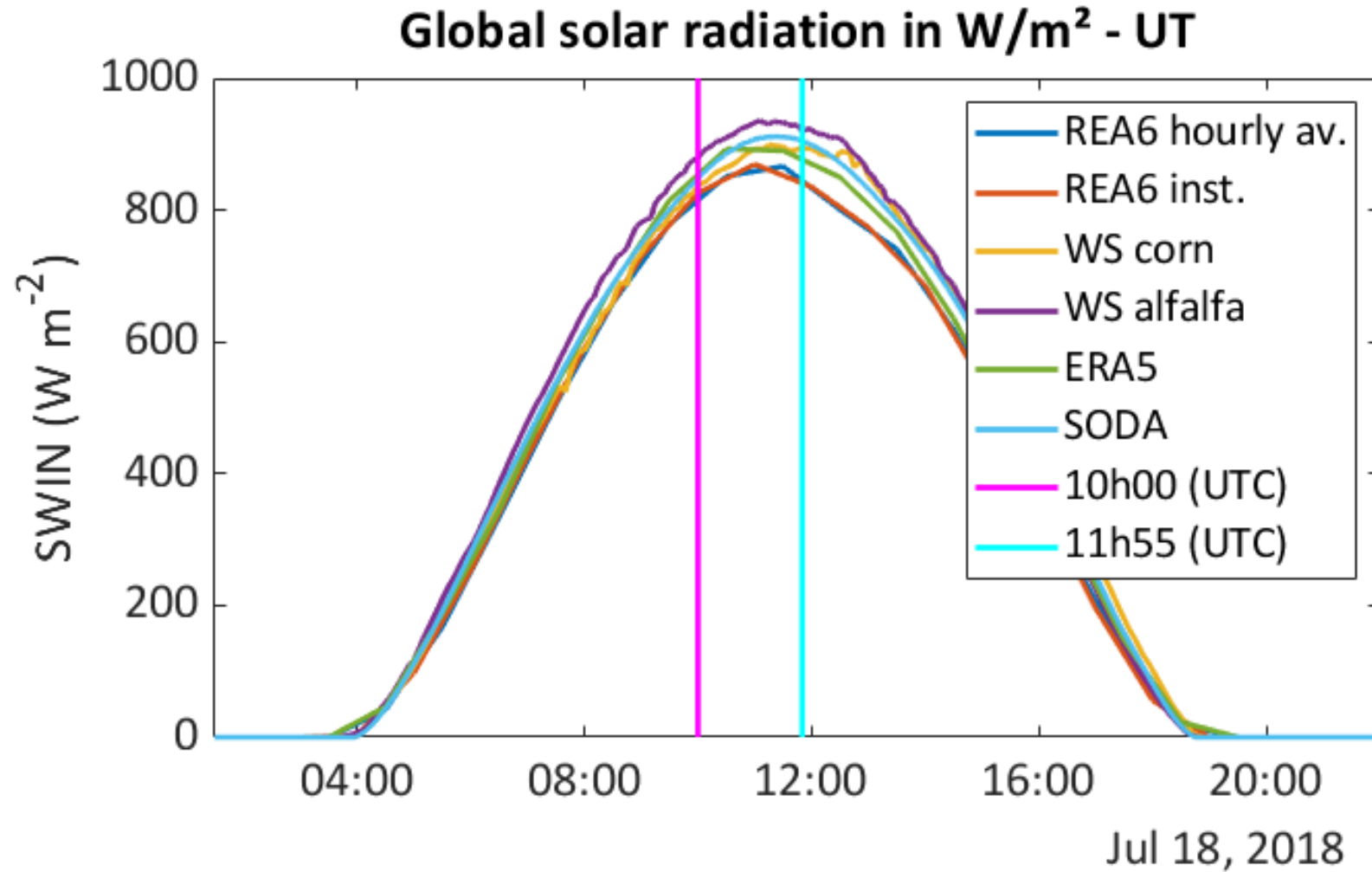
Air temperature and wind speed {
-> ERA 5
-> COSMO6-REA Reanalysis (Bonn University)
-> in situ data (2 locations)

Uncertainty analysis : we use the **EVASPA** concept (Gallego et al. 2013, Oliosio et al. 2018, Allies et al. 2020)

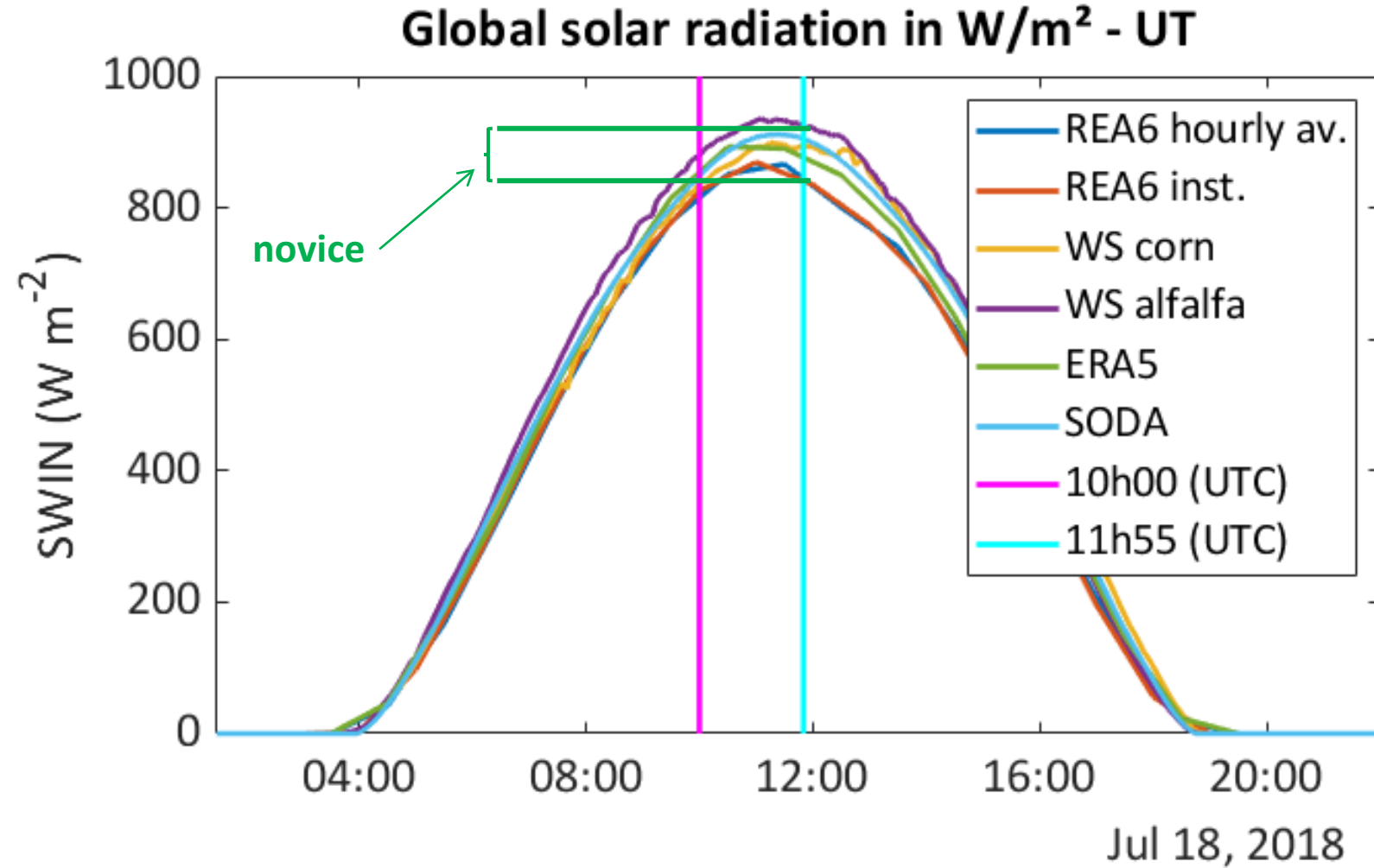
- > computes ET as an ensemble estimate (ensemble mean)
 - > from various sources of data
 - > by combining various model
- > computes uncertainties from the variability of calculated ET depending on
 - > model
 - > inputs

- > Uncertainty is defined as the standard deviation of the ensemble simulations (or the range of variations as in Mira et al., RSE 2016)
- > Uncertainty related to the impact of one variable or model is computed by keeping only the variations related to this model or variable
- > We compute two levels of uncertainty: (Allen et al., 2011, Blatchford et al., RSE 2019).
 - novice uncertainty = full range of variations
 - expert uncertainty = reduced range of variations depending on prior-information on accuracy

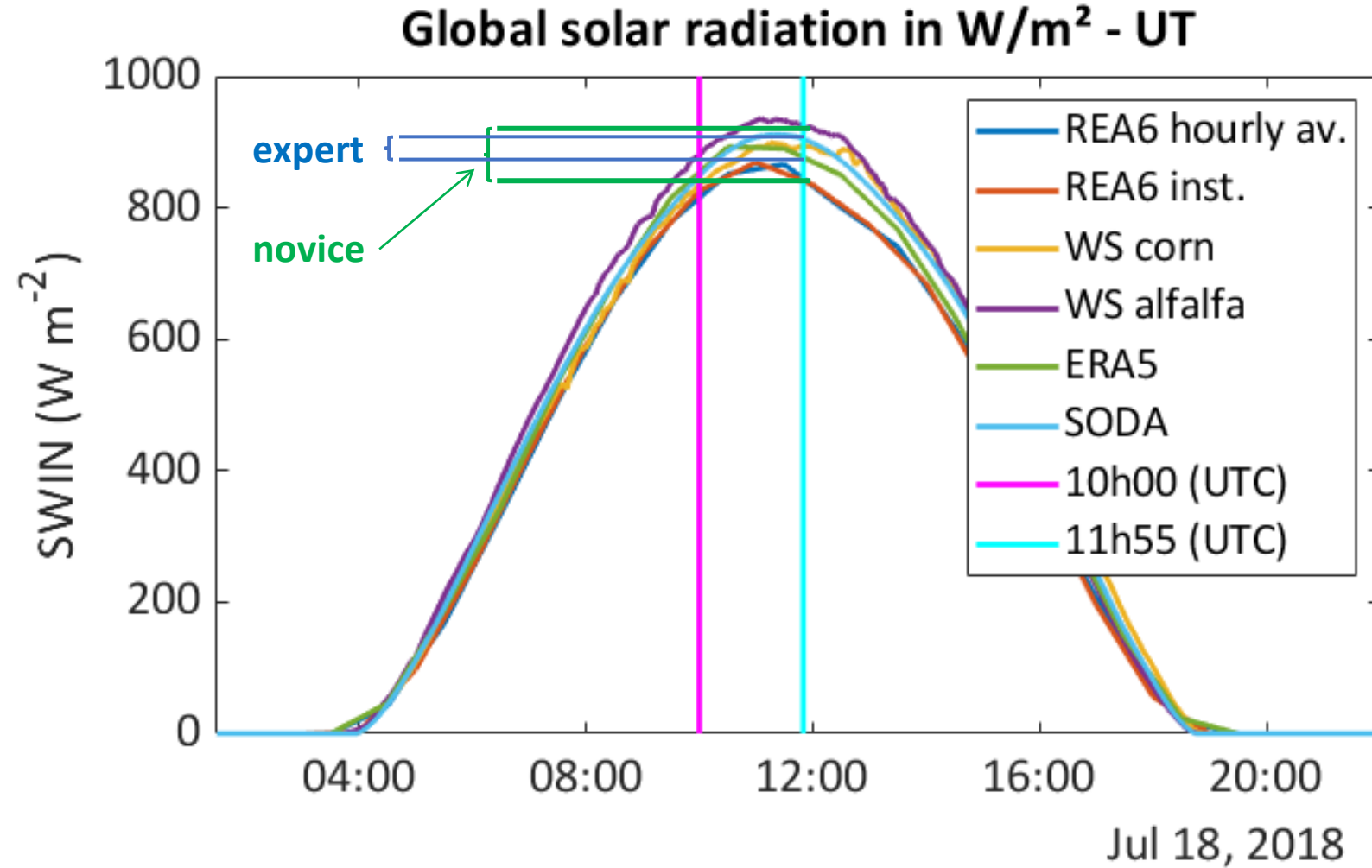
Example -> Meteorological variables : solar radiation



Example -> Meteorological variables : solar radiation



Example -> Meteorological variables : solar radiation



Uncertainties

variability in input data source and algorithms

	Novice case			Expert case			
common inputs							
Surface temperature	T_s	4 2 atmospheric corrections (Un. Valencia & CzechGlobe)					
	$NDVI$	2 calculations (different spectral windows)					
	$albedo$	13 models			11 models		
	$emissivity$	20 models			5 models		
Incident solar radiation	R_g	2 reanalysis	1 satellite product	2 in situ stations	1 reanalysis	1 satellite product	1 in situ station
Atmospheric radiation	$L \downarrow$	2 reanalysis		2 in situ stations	2 reanalysis		2 in situ stations
G to Rn ratio	$\xi (G/Rn)$	7 equations			6 equations		
Evaporative fraction model							
Evaporative fraction	EF	5 algorithms			5 algorithms with slope limitations		

Surface temperature data from TASI

Two sets of data were provided based on the TES algorithm

- one directly from ESA (CzechGlobe processing)
- three processing at University of Valencia :
 - > LIRE radiosoundings (on site)
 - > GSW profile (atmospheric model ...)
 - > NCEP profile (US model)

We used data on the 18th and the 20th at :

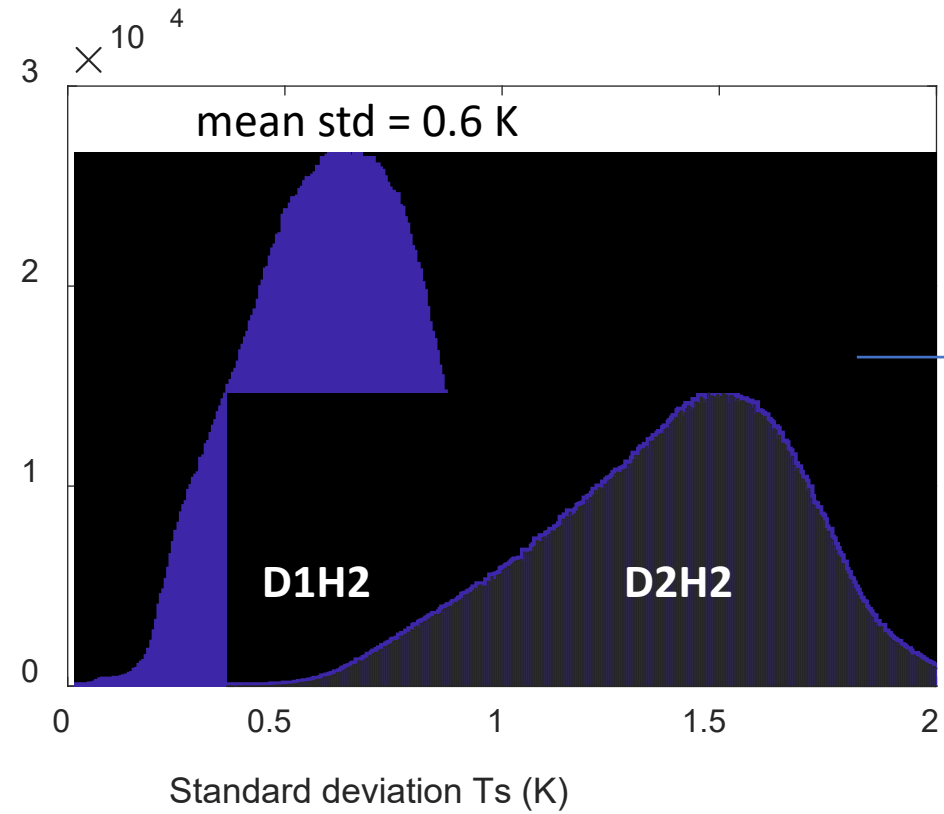
- around 10h UTC (= 12h CEST) -> morning
- around 12h UTC (= 14h CEST) -> midday

notation: D1H1, D1H2, D2H1, D2H2

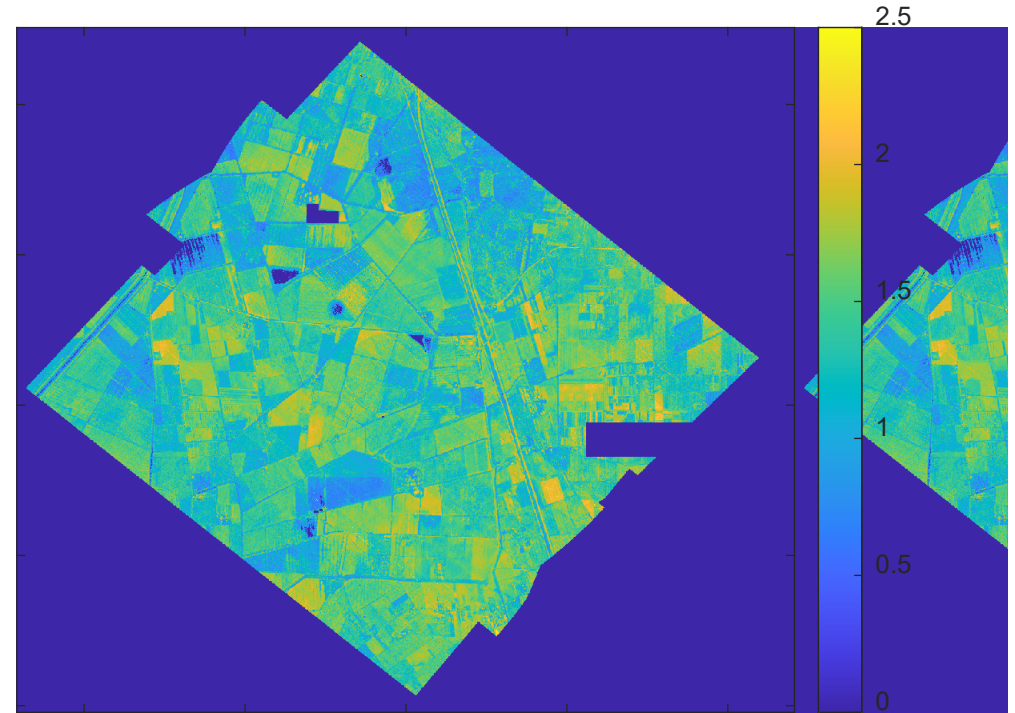
However, results were very similar for each days (and even each acquisition time) -> **D2H2**

Uncertainty in Ts

Standard deviation of the 4 Ts maps



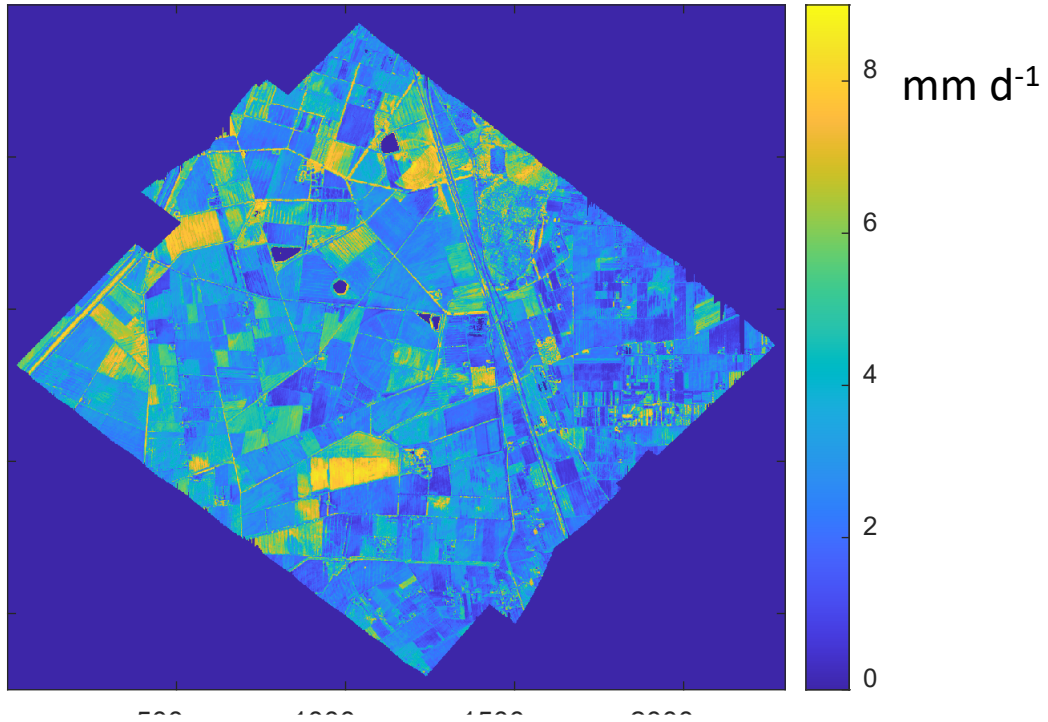
Ts standard deviation map D2H2 (K)



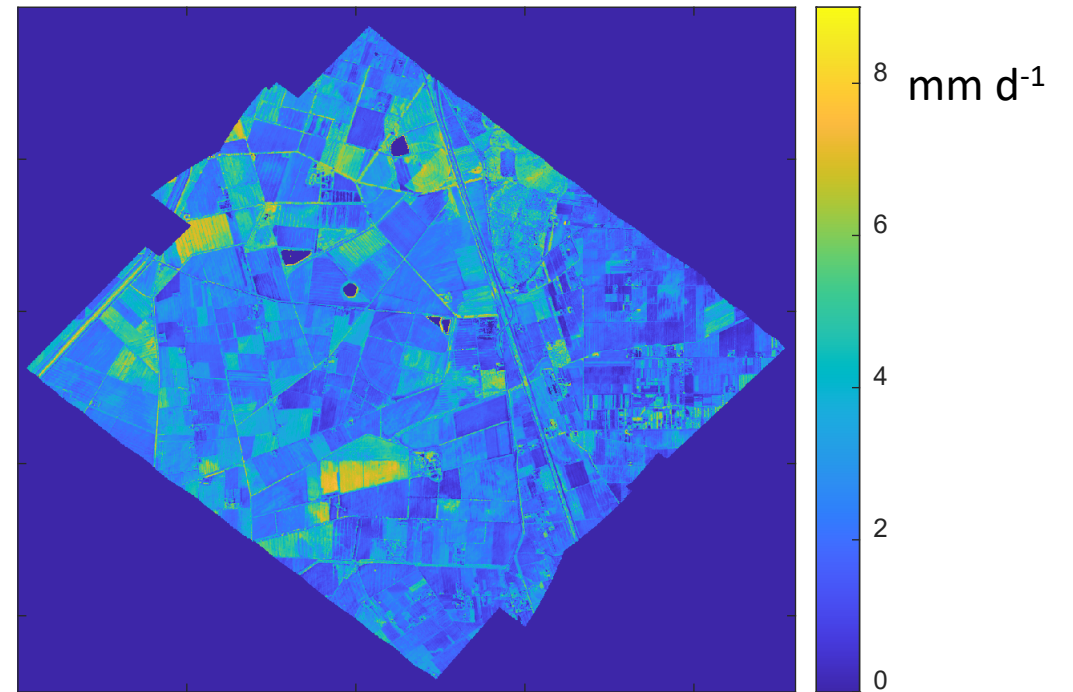
Evapotranspiration map

Average of all the calculations for D2H2

Novice case – average ET = 3.4 mm d⁻¹



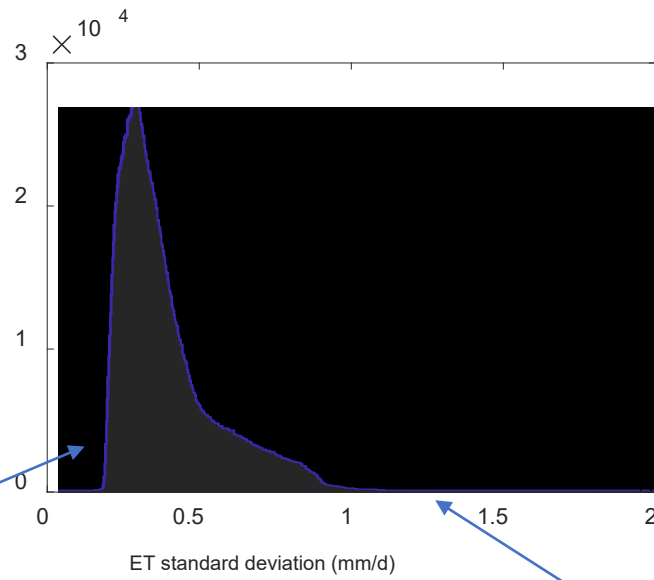
Expert case - average ET = 2.8 mm d⁻¹



Uncertainty in ET

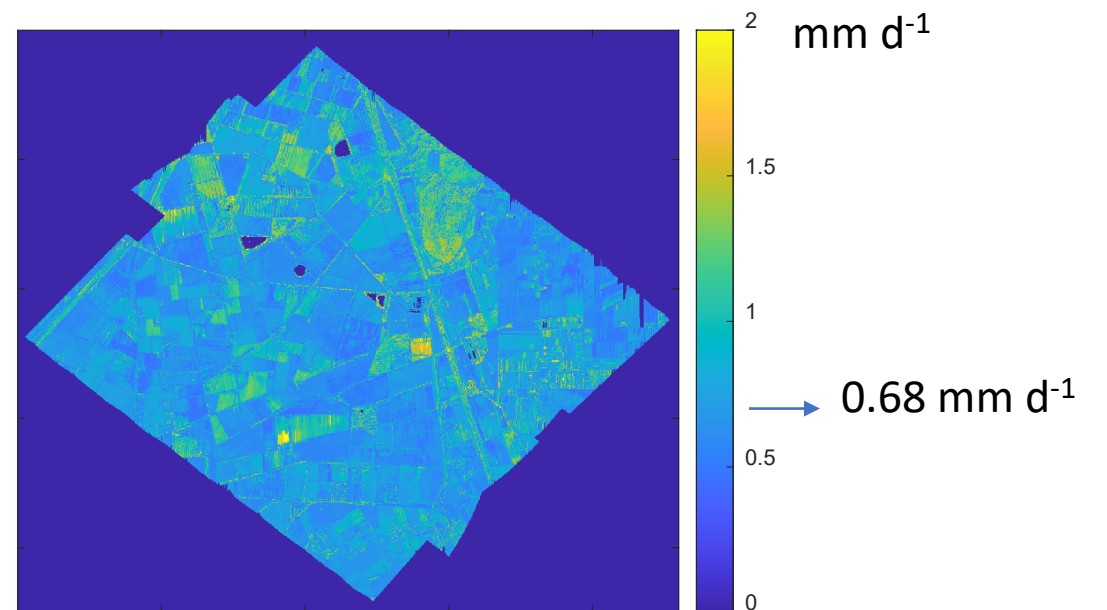
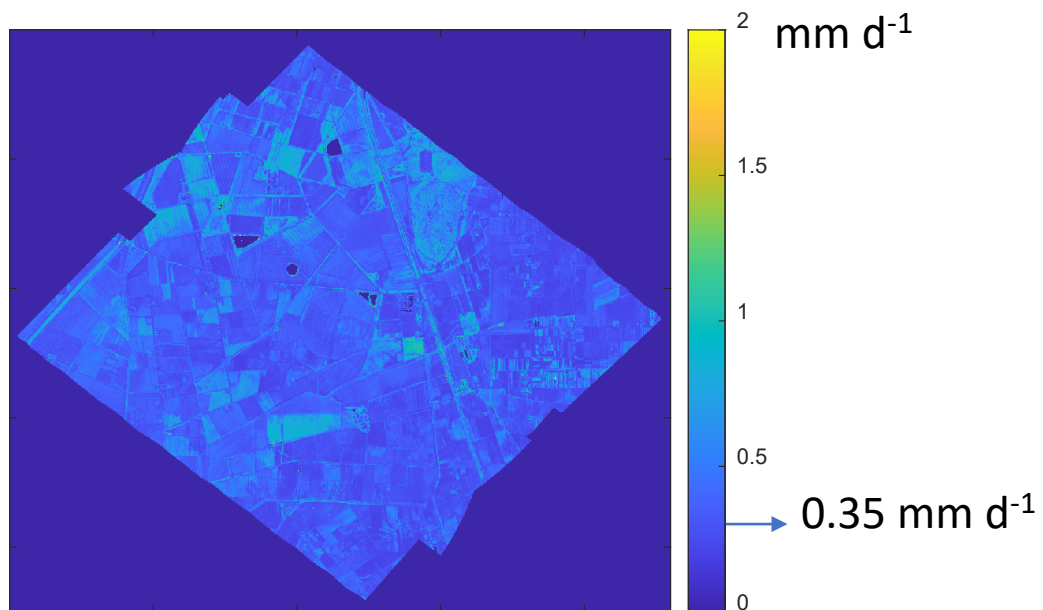
(standard deviation mm d^{-1})

D2H2 case



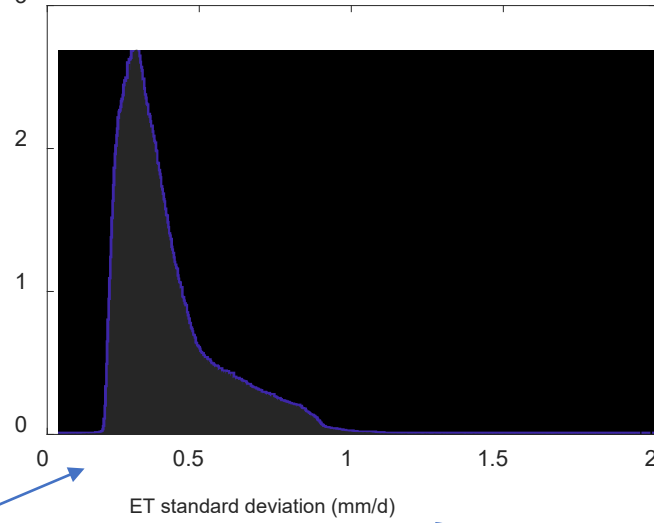
Expert case

Novice case

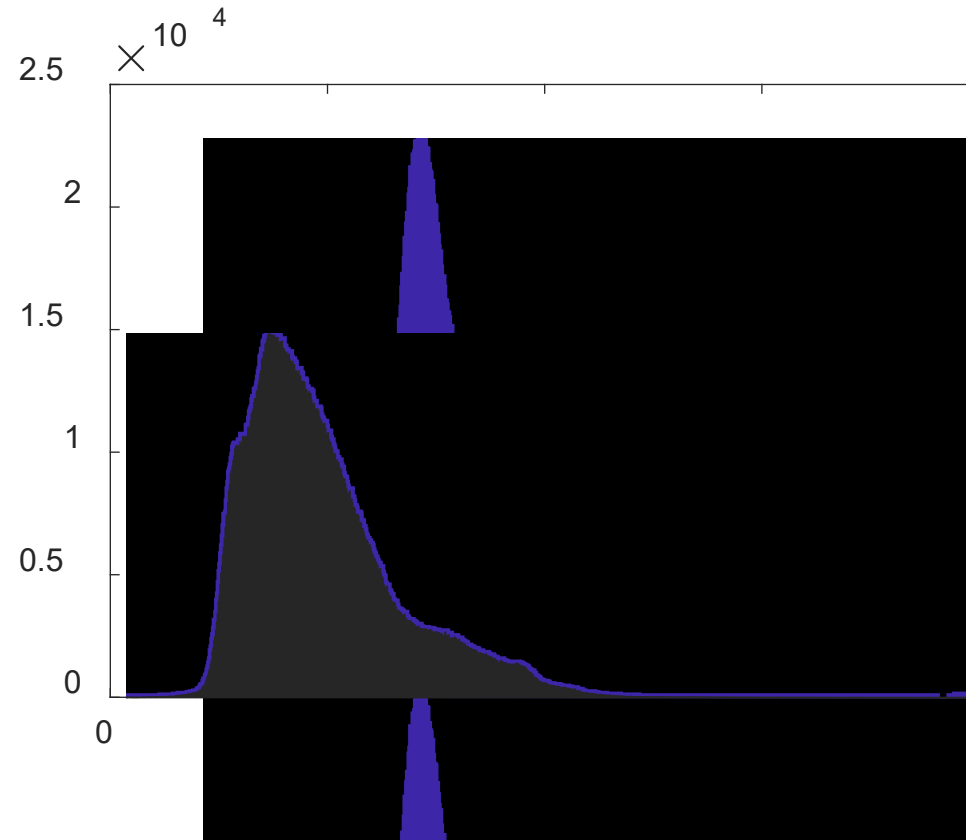
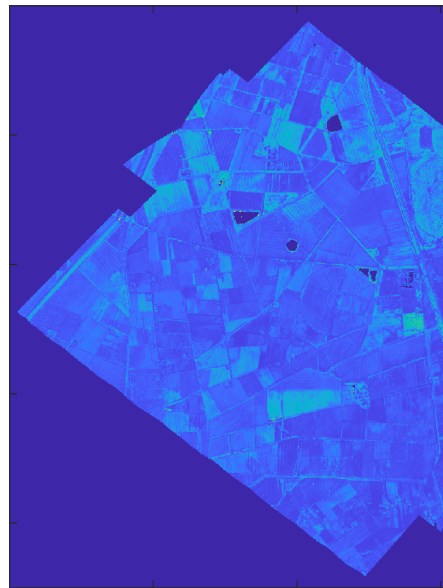


Uncertainty in ET
(standard deviation mm d^{-1})

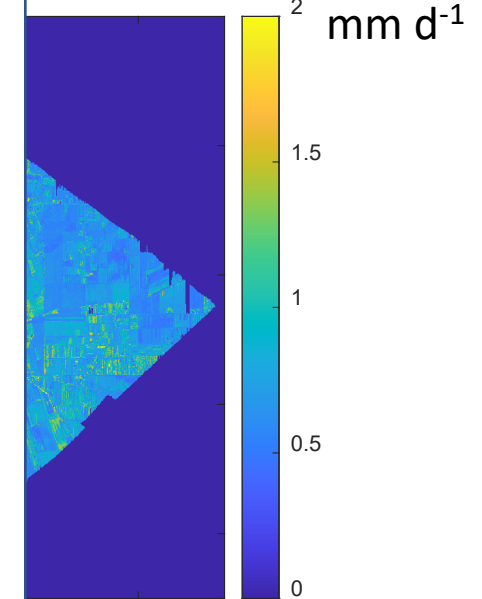
D2H2 case



Expert case

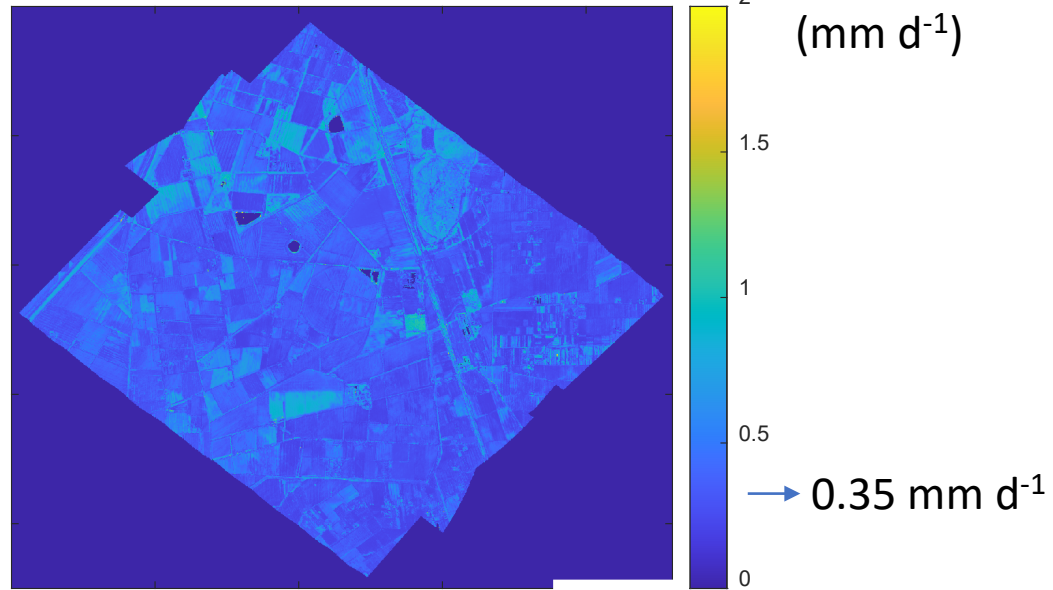


novice case

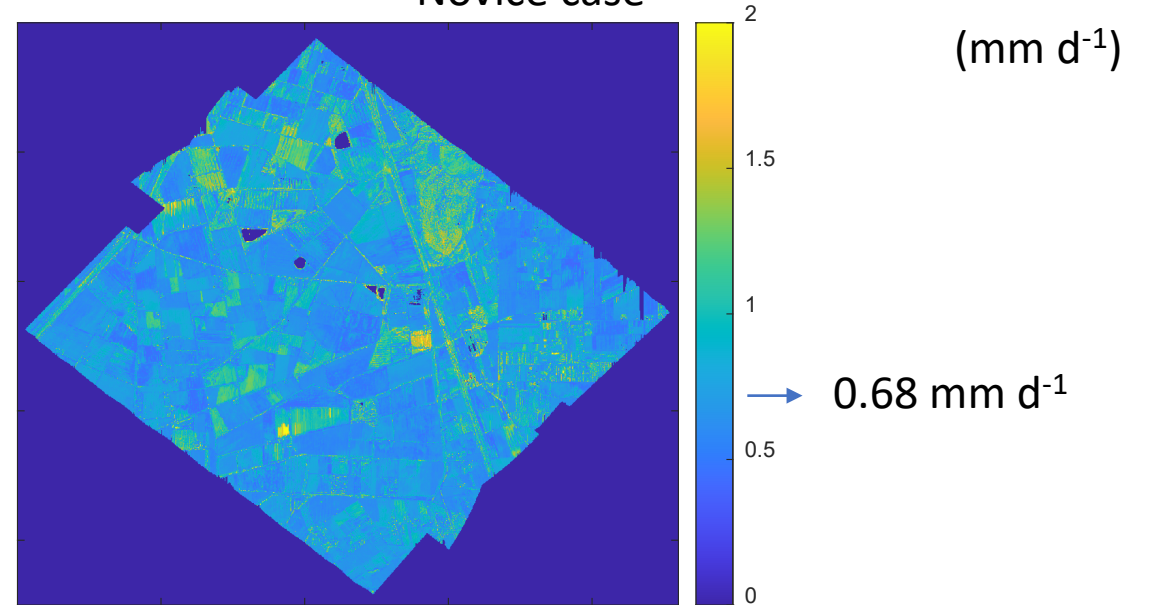


Uncertainty in ET (mm d⁻¹) D2H2 case

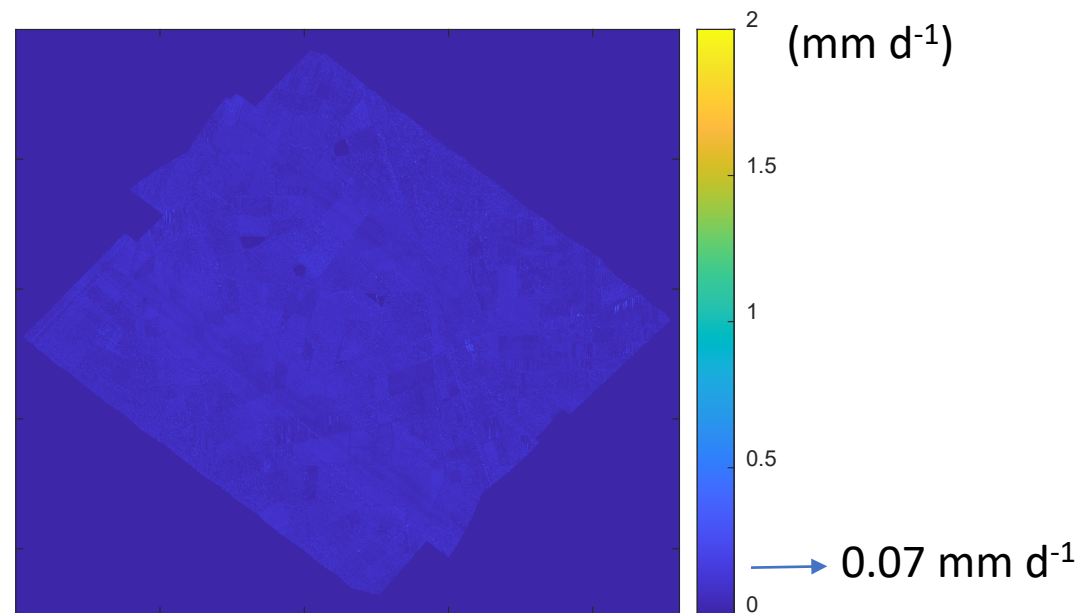
Expert case



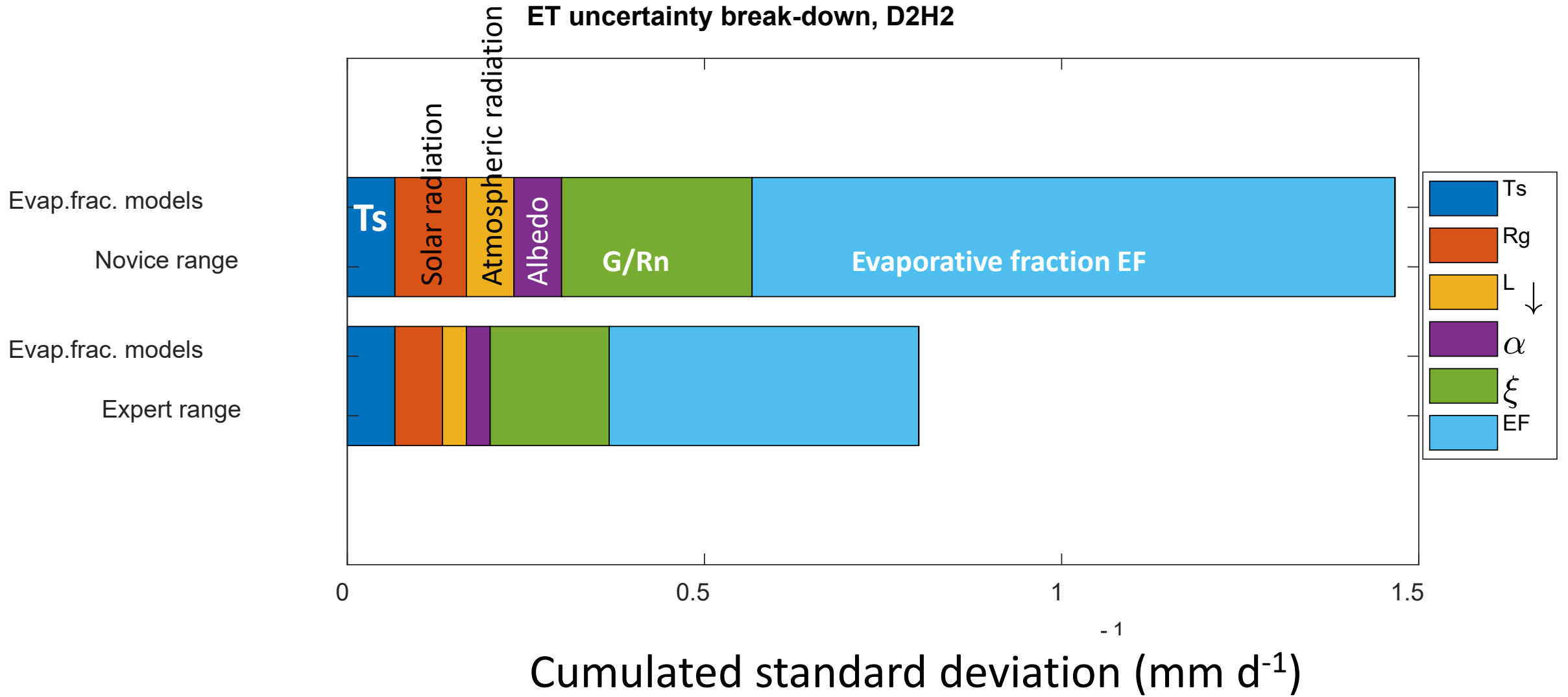
Novice case



Uncertainty related to Ts:



ET uncertainty break-down, D2H2



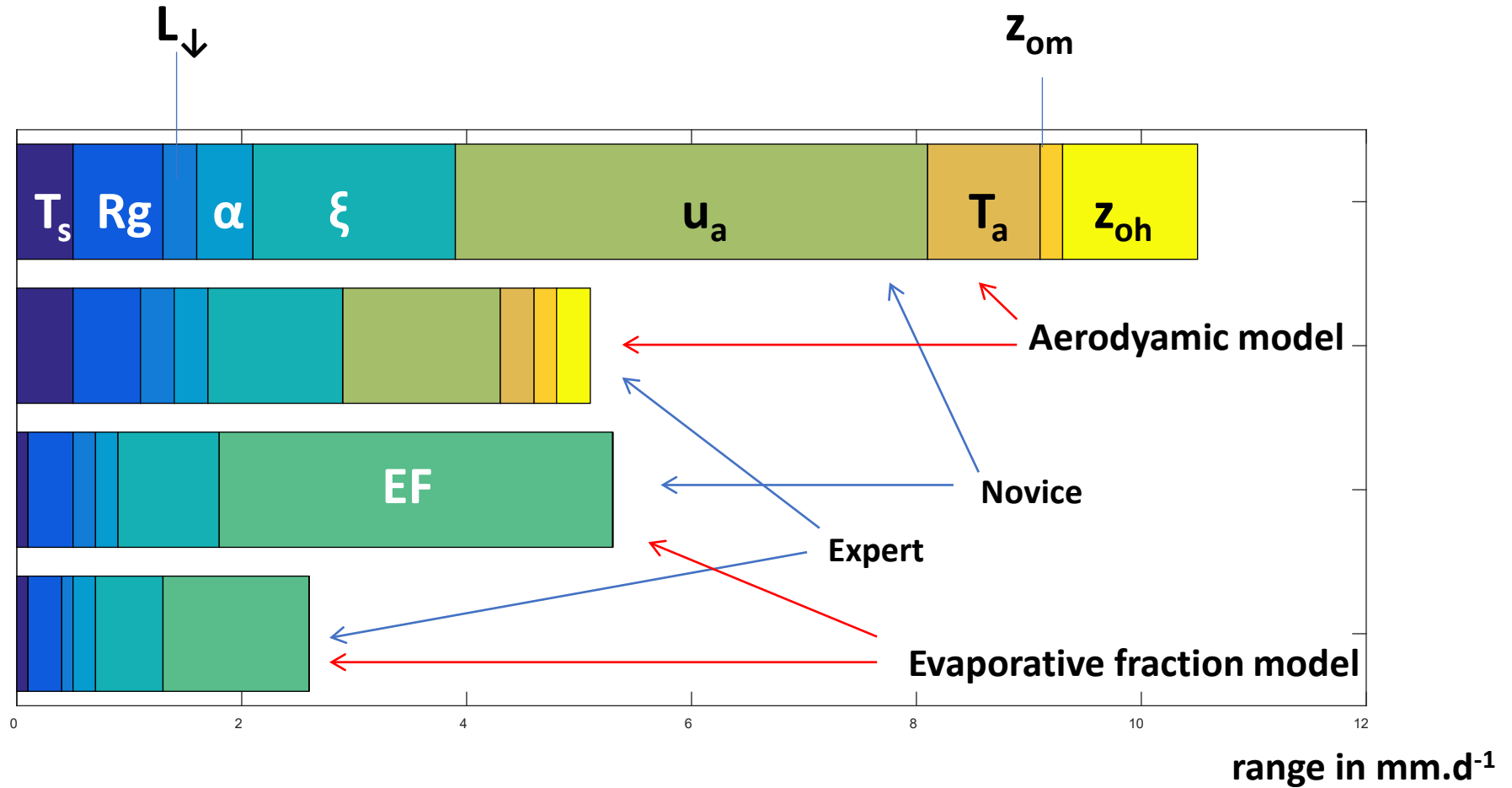
Summary

- uncertainty in ET is large, up to 1.5 mm.d^{-1} (for D2H2 and when expressed as standard deviation)
- uncertainty is significantly lower in the expert case than in the novice case
- ranking of uncertainty sources highlights
 - impact of T_s is low
 - largest impacts: evaporative fraction, G/R_n ratio (ξ)
 - model formulations have a larger impact than input data

Perspectives

- work on other situations
- extending to aerodynamic one-source and two source-models (TSEB, SPARSE...)
- transfer to time serie processing (ex. using MODIS data)
- derive an uncertainty algorithm to be used in the TRISHNA data processing for associating uncertainty to ET

Ranking of uncertainty sources



Uncertainty ranking for D2H1 (July 20th morning)