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**Editor decision: Publish subject to technical corrections**

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I am satisfied by the implemented changes by the authors according to the referees' requests. I therefore invite the authors to submit the final version of the manuscript.

We thank the editor for the positive decision on our manuscript.

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**Responses to referee comments**

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**Reviewer #1:**

Dear Referee#1,

We thank you very much for your valuable feedback and apologies for the delayed response as we had to wait for the editor's go-ahead before proceeding. Your suggestions have been very helpful. We have replied to your comments individually and revised the manuscript accordingly.

For your convenience, please see attached zipped file containing:

- Our responses (your comments are in black while our responses are in blue) :  
*2025-4522\_ResponsestoRC1.pdf*
- Revised manuscript with tracked changes :  
*2025-4522\_RevisedManuscript\_withTC.pdf*
- Revised supplementary document with tracked changes :  
*2025-4522\_RevisedSupplement\_withTC.pdf*

Thank you again.

Sincerely,

Mwangi et al.

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**Referee Comments (and our responses):**

<https://doi.org/10.5194/egusphere-2025-4522-RC1> ; <https://doi.org/10.5194/egusphere-2025-4522-AC1>

**General assessment**

This manuscript presents a scientifically robust and timely contribution. The ensemble-based EVASPA framework, combining multiple LST, radiation, EF and G datasets, provides a valuable foundation for analysing uncertainties in contextual evapotranspiration (ET)

modelling. The methodological implementation is technically sound, and the topic is highly relevant for the remote-sensing and ET communities, especially in the context of upcoming missions such as TRISHNA.

My comments concern almost exclusively the readability and organisation of the manuscript, not the scientific validity, which appears strong. The manuscript is dense, and the large amount of information sometimes makes it challenging to identify the central messages. A clearer narrative and more guidance for the reader would significantly increase the accessibility and impact of the work.

We greatly appreciate your taking the time to read our manuscript and providing valuable feedback. Thank you. Please see our comments below and the modifications in the revised manuscript addressing your concerns.

## Major comments

### 1. Readability and narrative flow

The manuscript contains extensive information, often presented in long paragraphs with multiple embedded ideas. This makes it difficult for readers to extract the main points and to follow the progression of the results. A clearer hierarchy of information, distinguishing essential findings from detailed descriptions, would be highly beneficial.

Thank you for the suggestion. We have made adequate improvements to the manuscript for better readability. For the modifications made, kindly refer to the revised manuscript with tracked changes.

Example: In parts of Section 4.3, very long multi-clause sentences make it difficult to isolate the key conclusions.

Well noted. The section has been refined. We have shortened the long sentences in this section (as well as elsewhere in the manuscript). Please refer to the revised manuscript (with TC) for the corrections made.

### 2. Clarity of objectives and role of each analysis

The study combines input-data variability analysis, performance evaluation against flux towers, ensemble-based uncertainty quantification, and similarity clustering. All of these components are relevant, but the manuscript does not always clearly articulate how each one contributes to the overarching objective. Briefly restating the purpose of each major section would help maintain coherence.

We appreciate your feedback. The objectives have been made clearer.

### L99-107:

“The objectives of this study are therefore to: 1) preliminarily assess the temporally continuous ensemble estimates of contextual ET in preparation of the TRISHNA SWEAT products, and 2) quantify and analyse the inherent uncertainties introduced by various factors/variables in long term contextual ET estimates. The performance assessments are

carried out against in-situ data to ensure consistency and provide a reference for comparison. Validation is however not the primary focus of this work as the evaluation of EVASPA algorithms is part of the currently ongoing TRISHNA ET benchmarking activities (particularly with high spatial resolution data). Here, we mainly seek to systematically identify and analyse the principal sources of uncertainty inherent in the model, arising from both input data variability and methodological assumptions. An analysis of the uncertainty in the EVASPA members relative to the four parameters/variables (i.e., LST, radiation, EF and G flux methods) is thus presented.”

3. Interpretation and introduction of figures

The figures are well designed, but several require more explanation to be fully interpretable. In some cases, it is not clear how values were aggregated or what specific elements represent.

More details have been added on how the figures were constructed and from which data.

E.g. Table 2

Autumn	0.70	0.66	-0.05	0.54	0.30	0.34	0.24	0.83	0.60	0.07	0.65	0.34	0.31	0.22
All sites and seasons combined	0.94	0.77	-0.09	0.69	0.40	0.34	0.23	1.16	0.73	0.03	0.87	0.47	0.31	0.22

\* Reported uncertainty are the median values over the uncertainty time series';  
E.g. for SD, the standard deviation is calculated for the ensemble at each time step, then the median of the SD time series is reported.

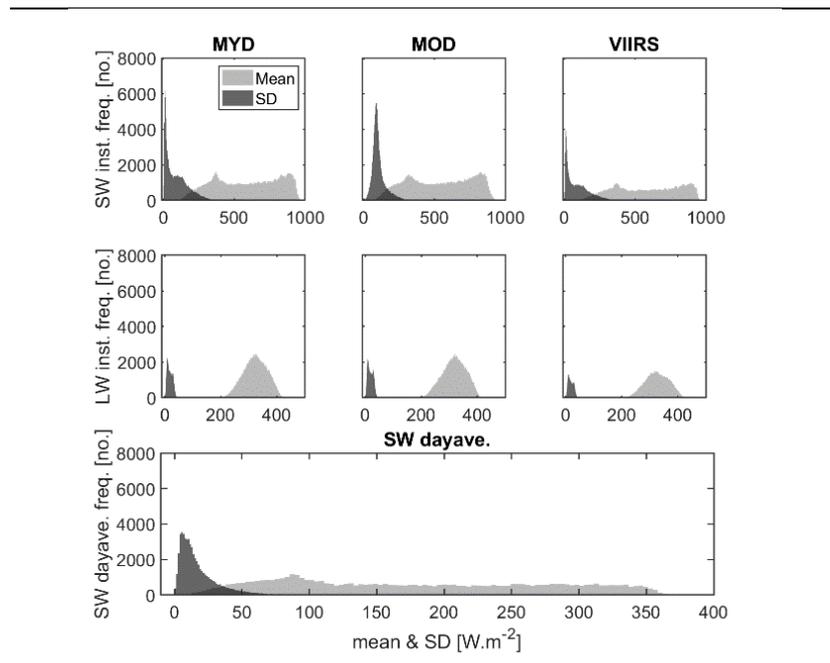
Examples:

- For Figure 3, the method used to compute the distributions (means and standard deviations) is not entirely clear, particularly the number of pixels and temporal samples considered.

The figure caption has been updated.

**L386-389:**

“



**Figure 1: Global uncertainty in the radiation input data over the eight study sites – one pixel per site over the 2004-2024 study period. a) shortwave (SW) radiation at the three unique satellite overpasses: i.e. AQUA (MYD), TERRA (MOD), and Suomi NPP (VIIRS), respectively; b) same as a) but for longwave (LW) radiation; c) daily shortwave radiation - applies for all satellites / sensors.**

”

- Figure 5 would benefit from a clearer explanation of what each column corresponds to in terms of ensemble subsets.

We have added more information on the ensemble estimates used to compute the performances shown in the columns. The figure caption as well as its description in the text and appendix have been updated.

**L474-476:**

“Figure 5: Performance heat maps (RMSD, D, and Bias) for the different EVASPA models across all evaluation sites. The EVASPA models are composed of all gap-filled member estimates in the set (e.g. the MYD11 set is constructed using the MYD11 LST/E data together with the 3 radiation datasets, 9 EF and 9 G methods). More details on how the sets are constructed are contained in Figure S 1 in the supplements.”

**Supplements:**

“

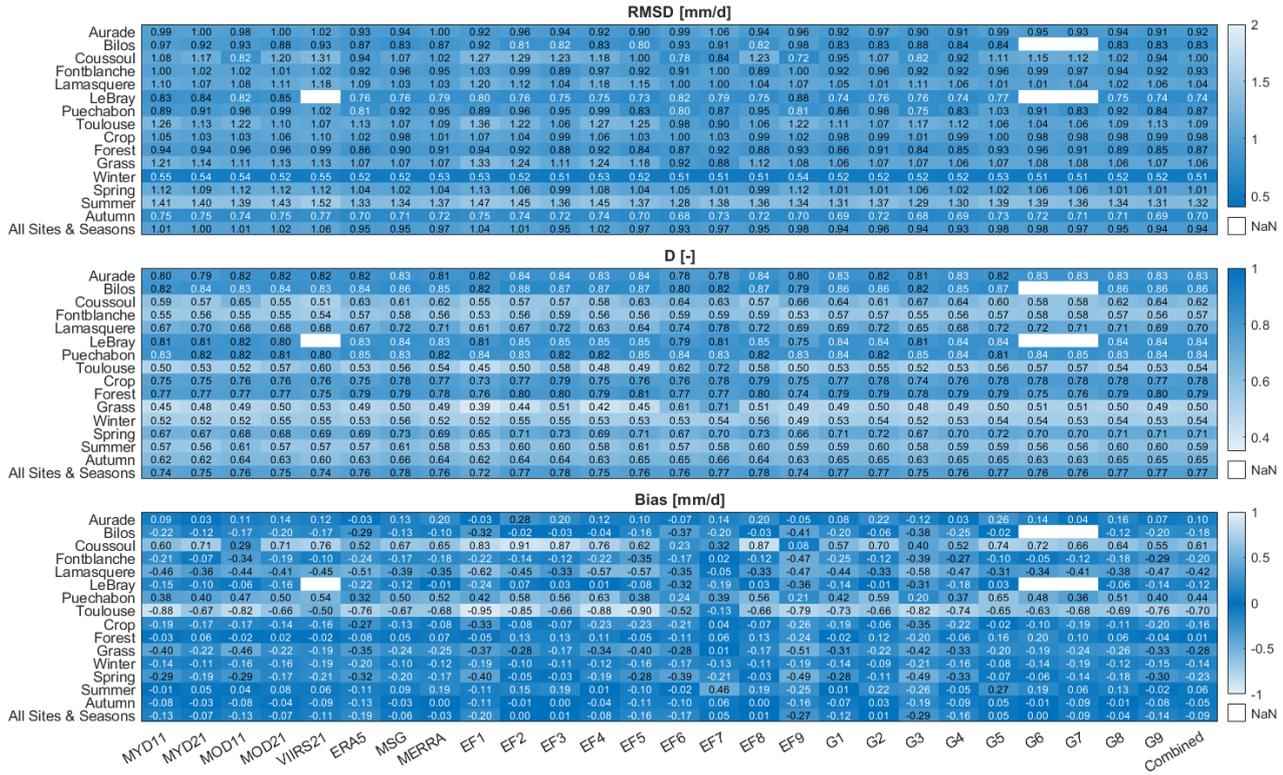


Figure S 1: As in Figure 5, but including VIIRS-based set\* estimates.

\* EVASPA model / set construction:

- *LSTs example: the MYD11 set is constructed taking MYD11 LST/E as the unique LST dataset used together with the 3 radiation datasets, 9 EF and 9 G methods for 243 estimates.*
- *Radiation example: the ERA5 set is constructed taking ERA5 as the unique radiation dataset used together with the 4 LSTs (or 5 LSTs here, including VIIRS), 9 EF and 9 G methods for 324 (or 405 here, including VIIRS-based) ET estimates.*
- *EFs example: the EF1 set is constructed taking EF1 as the unique EF method used together with the 4 LSTs (or 5 LSTs here, including VIIRS), 3 radiation datasets, and 9 G methods for 108 (or 135 here, including VIIRS-based) estimates.*
- *G example: the G1 set is constructed taking G1 as the unique G flux method used together with the 4 LSTs (or 5 LSTs here, including VIIRS), 3 radiation datasets, and 9 EF methods for 108 (or 135 here, including VIIRS-based) estimates.*

”

Providing more explicit introductions to figures would greatly help readers understand what to focus on.

Thank you for the feedback. We have updated the figure descriptions and captions accordingly. Kindly refer to the revised manuscript with tracked changes.

#### 4. Spatial representativeness

Most analyses are based on the eight 1-km pixels corresponding to the flux-tower sites. This setup is fully appropriate for a site-based uncertainty assessment, but it does limit spatial generalisation. A brief acknowledgement of this limitation would help clarify the scope of the conclusions.

Well noted, thank you. There is indeed a spatial limitation. We have added text acknowledging the spatial generalization limitation.

**L696-704:**

“Whereas this study is generally appropriate for site-based uncertainty assessments, we also acknowledge that several spatial scale issues exist that limit spatial generalization and transferability of the results. In particular, there is a clear scale mismatch between the estimates derived from (and for) the MODIS pixel, which represent relatively large and heterogeneous spatial areas, and the in-situ observations, which are representative only of the localized EC flux tower footprint. As such, the performance assessment results should be interpreted with caution, as the comparison involves estimates and observations with fundamentally different spatial coverage and representativeness. This limitation is further compounded by the presence of mixed pixels, where multiple land cover types and surface characteristics are aggregated within a single pixel, introducing additional uncertainty that is not explicitly accounted for in our analyses.”

**5. Emphasis on key insights**

Some of the most important conclusions - such as the dominant role of LST (particularly overpass time), the notable influence of EF methods, and the different impact of radiation in gap-filling versus instantaneous estimates - are present but sometimes buried within dense text. Highlighting these insights more explicitly would strengthen the communication of the study's main contributions.

We have updated the text to clearly highlight the key insights from our study. Please see our comments below and the various modifications within the revised manuscript.

**L668-718:**

“

...

Practically, our findings suggest that:

*a)* Even when using a simple ensemble average, performance over the long-term is generally reasonable across all sites. However, agreement appears to diminish especially over heterogeneous pixels. Performance of the individual member estimates within the ensemble also varies significantly, with the different constituent input and methods showing varying performance depending on the site. For example, the TES-derived LSTs appear to consistently outperform the GSW-derived LSTs over the Toulouse grassland site. As expected, the absolute seasonal performance (i.e. RMSD, MAE) is primarily influenced by the availability of radiation.

*b)* Of all the four considered variables, the LST inputs introduce the largest spread (thus uncertainty in terms of SD) in the EVASPA ETs, followed by the EF methods and the radiation products, with the ground heat flux methods – G - having the least influence. This highlights the key role of LST and EF methods in determining:

- The available energy ( $R_n - G$ ) at the near-land surface (where LST is a key input term for the net radiation,  $R_n$ ).
- How this available energy is partitioned between the turbulent heat fluxes in contextual modelling of ET, i.e., when deriving the evaporative fraction at the pixel scale.

While the differences in temperature-emissivity separation methods impact the agreement between ensemble members, the temporal aspect of LST/E acquisition introduces a broader and more systematic influence on ET uncertainty – primarily due to the associated variability in the radiation inputs at those times. For applications requiring high precision or sensitivity to temporal dynamics (e.g., crop stress monitoring or irrigation scheduling), these timing effects may need to be accounted for explicitly, either through ensemble weighting strategies or temporal normalization approaches. Future ET modelling efforts may benefit from leveraging terrestrial emission data from missions with higher revisits so as to minimize the uncertainty associated with acquisition time.

...

”

### **Minor comments**

- Some sentences are particularly long and could be split for clarity.

We have shortened (and clarified) the long sentences throughout the manuscript. Please see the revised manuscript with tracked changes for the modifications made.

- Acronyms may be reintroduced when they reappear after long intervals.

Text updated accordingly; we have recalled description of (non-standard) acronyms after long intervals

- The introduction could converge more directly toward the specific objectives of the study.

Thank you. We have updated the text accordingly.

### **L99-101:**

“... The objectives of this study are therefore to: 1) preliminarily assess the temporally continuous ensemble estimates of contextual ET in preparation of the TRISHNA SWEAT products, and 2) quantify and analyse the inherent uncertainties introduced by various factors/variables in long term contextual ET estimates.”

- The conclusion is informative, but a more concise synthesis of the central messages would improve its effectiveness.

We appreciate the reviewer’s comment. The conclusion has been updated making it clearer.

**L668-683:**

“... our findings suggest that:

*a)* Even when using a simple ensemble average, performance over the long-term is generally reasonable across all sites.

...

*b)* Of all the four considered variables, the LST inputs introduce the largest spread (thus uncertainty in terms of SD) in the EVASPA ETs, followed by the EF methods and the radiation products, with the ground heat flux methods – G - having the least influence.

...”

**Final recommendation**

The science is solid and the manuscript has clear potential. Once clarified and streamlined, it will make a strong contribution. I recommend major revisions, focusing mainly on the readability, structure, and presentation rather than on the scientific methodology.

Again, many thanks for your review and feedback. In addition to our responses above, please see the revised manuscript (and supplements) with tracked changes.

**Reviewer #2:**

Dear Referee#2,

We greatly appreciate your valuable comments, which have helped us improve our manuscript extensively. We have replied to your comments individually and revised the manuscript accordingly.

For your convenience, please see attached zipped file containing:

- Our responses (your comments are in black while our responses are in blue):  
*2025-4522\_ResponsestoRC2.pdf*
- Revised manuscript with tracked changes :  
*2025-4522\_RevisedManuscript\_withTC.pdf*
- Revised supplementary document with tracked changes :  
*2025-4522\_RevisedSupplement\_withTC.pdf*

Thank you very much.

Sincerely,

Mwangi et al.

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**Referee Comments (and our responses):**

<https://doi.org/10.5194/egusphere-2025-4522-RC2> ; <https://doi.org/10.5194/egusphere-2025-4522-AC2>

This manuscript presents a comprehensive analysis of uncertainties in long-term ensemble estimates of contextual evapotranspiration (ET) over Southern France using the EVASPA modelling framework and MODIS remote sensing products. The study addresses an important and timely topic, as uncertainty quantification in satellite-based ET estimates remains a major challenge for hydrological and climate applications.

The ensemble-based approach adopted by the authors is methodologically sound and well aligned with current best practices in remote sensing of land–atmosphere fluxes. The manuscript demonstrates a strong technical foundation, and the comparison with flux tower observations adds credibility to the analysis.

Thank you very much for taking the time to read our work and providing valuable feedback. We have addressed your concerns accordingly; please see our comments below and the modifications within the revised manuscript.

**Major comments**

- A key issue is the comparison against the daily ET estimates measured by flux towers. Is their footprint being compared with the pixel dimension? A scale issue should always be addressed when discussing such comparisons.

We do agree that there is a likely mismatch between the tower footprint and the pixel size. We have added text acknowledging the scale issue introduced when ET estimates are compared against in-situ EC data.

**L696-704:**

“Whereas this study is generally appropriate for site-based uncertainty assessments, we also acknowledge that several spatial scale issues exist that limit spatial generalization and transferability of the results. In particular, there is a clear scale mismatch between the estimates derived from (and for) the MODIS pixel, which represent relatively large and heterogeneous spatial areas, and the in-situ observations, which are representative only of the localized EC flux tower footprint. As such, the performance assessment results should be interpreted with caution, as the comparison involves estimates and observations with fundamentally different spatial coverage and representativeness. This limitation is further compounded by the presence of mixed pixels, where multiple land cover types and surface characteristics are aggregated within a single pixel, introducing additional uncertainty that is not explicitly accounted for in our analyses.”

- Furthermore, even though the main goal here would be the spread of the “error”, more than the error itself, it would be helpful to sort some results in terms of land cover, as you did with the four seasons, in order to assess potential for uncertainty in different environments.

In addition to seasonal analyses, we have added results comparing results in terms of land cover. Thank you for the suggestion.

**L560-563:**

“The heat maps corresponding to the different land covers and seasons are presented in the supplementary document. Across the land covers, the trends are generally similar to those in Figure 9, Figure 8 with only subtle differences (Figure S 4, supplements).”

**L618-626:**

“Only the bar graphs for all sites combined are shown - the site-specific graphs show a similar trend (see Figure A 2). From the plots, the overall trends (particularly on ranking of the four variables) appeared largely similar across the analyses, even when alternative reference combinations of the distinct variables were considered. This suggests that the observed patterns are largely consistent and not overly sensitive to the specific choice of reference inputs-methods combination. LST still appears to introduce most of the overall (all seasons) uncertainty over the three land covers (see supplements, Figures S 6, S 7 and S 8). Nonetheless, the salience of EF methods on the uncertainty over the grasslands and croplands is apparent, especially in the non-interpolated daily ET estimates during summer.”

**Supplements:**

In addition to Figures S 6, S 7, S 8 recalled below, also see Figure S 4, where the heatmaps are separately generated according to land cover type.

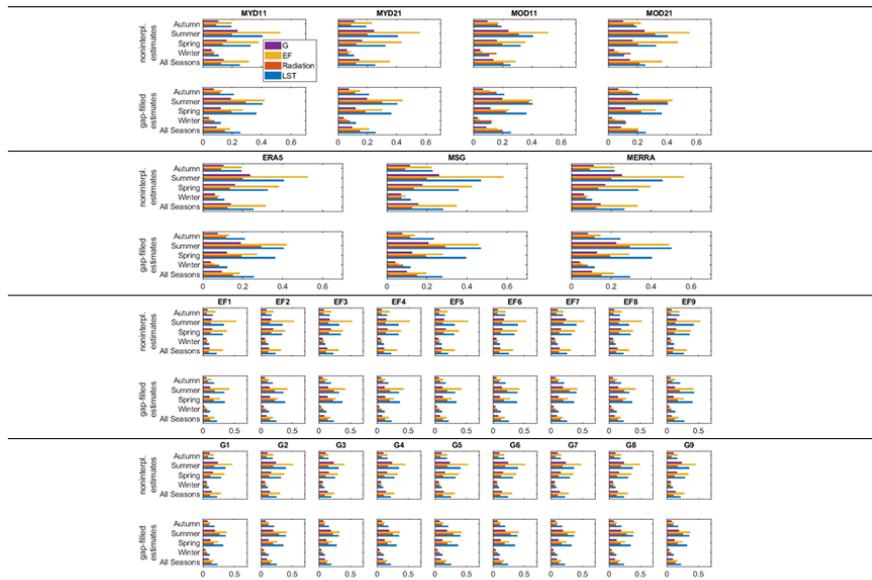


Figure S 6: As in Figure 12, but for Croplands – Auradé and Lamasquère.

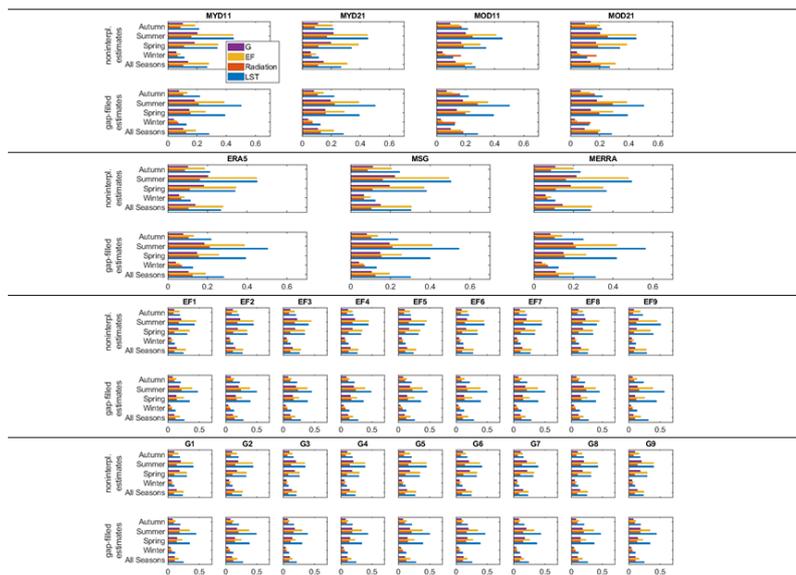


Figure S 7: As in Figure 12, but for Forests – Bilos, Le Bray, Fontblanche, Puchabon.

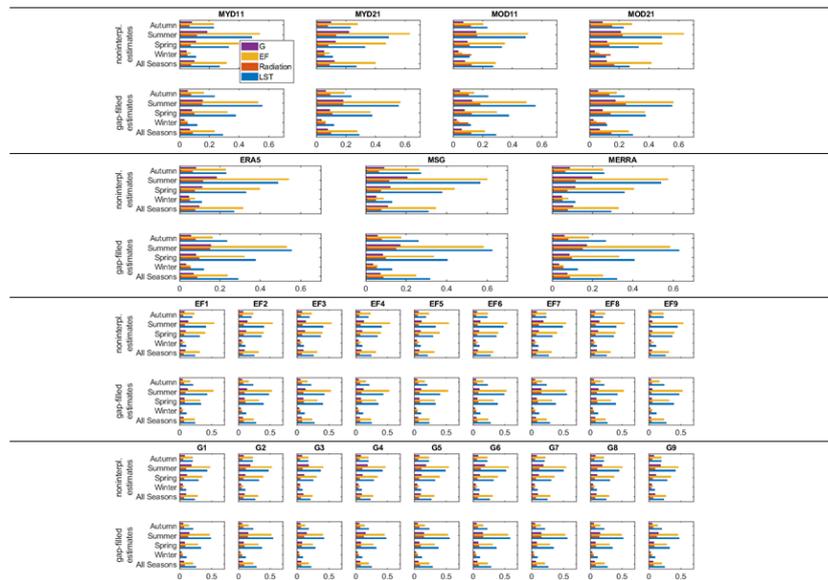


Figure S 8: As in Figure 12, but for Grasslands: **Caussoul, Toulouse.**

- L185-186. This assumption is quite common throughout the literature. However, it would be helpful to recall somewhere what uncertainties lie behind it, and, if possible, try to include them in the uncertainty quantification you obtain later from your ensembles

Thank you for the clarification. Indeed, it is a common assumption that  $\lambda E$ -to-solar radiation ratio is conserved throughout the day, which introduces further uncertainties in the estimated daily evapotranspiration estimates. We have updated and clarified the text accordingly to acknowledge such limitations.

### L203-216:

“Eventually, the daily  $ET_d$  ( $\lambda E/\lambda$ ) [mm/d] is computed using the daily solar radiation data, where it is commonly assumed that the  $\lambda E$ -to-solar radiation ratio is conserved throughout the day, i.e.  $\frac{\lambda E_1}{SW_1} \approx \frac{\lambda E_d}{SW_d}$  (e.g., Delogu et al., 2012).”

“It should be noted this is one of several methods used to aggregate latent heat fluxes to the daily scale. It was selected herein as it is physically based, requires minimal data and relatively simple to apply. We nonetheless acknowledge that it does introduce uncertainties in the estimations, especially since the assumption that the ratio is conserved throughout the day may not always hold. Within the TRISHNA SWEAT framework, other methods (e.g. conservation of the evaporative fraction throughout the day) are being tested to quantify such uncertainties between methods, and for possible inclusion as part of the product chain.”

- Since the focus here is uncertainty, I would have liked more quantitative references to the standard deviations of the different combinations (e.g., in Fig.5 it seems a missing information), both in absolute and relative (to the main variable values) terms. Furthermore, as seen in Fig.6, a lot of the ensemble results seems to distribute quasi-normally, although a non-negligible quantity (e.g., Aurade, Caussoul, LeBray, Toulouse) do not. Given these distributions, a skewness indicator or other, more complete parameters would seem to be better fit to describe the results of the different indexes you computed, as opposed to the simple SD, which can be misleading when the sample is not distributed normally.

Well noted. We acknowledge that standard deviation (SD) can be misleading/misinterpreted when used with non-Gaussian data. We however preferred using standard deviation throughout the main text since SD will be used as the primary uncertainty measure in the ensemble estimates within TRISHNA. Note that the interquartile range (IQR), which is useful in quantifying non-normal/non-symmetrical spreads, is already qualitatively depicted in the box plots displayed within the violin plots, Figure 6. We have now added the quartile coefficient of dispersion (and the coefficient of variation) to complement the SD.

**L235-240:**

“To be consistent with recent ensemble ET products, which quantify uncertainty in SD, the TRISHNA ET products will also include the ensemble SD as the primary uncertainty measure (this may change as the products evolve). We nonetheless acknowledge that SD (and the corresponding coefficient of variation, CV) can be misleading, especially when the analysed data exhibits non-symmetrical (specifically non-Gaussian) distribution. As such, the coefficient of quartile variation or quartile coefficient of dispersion (QCD)—which was recommended for non-normal distributions by Bonett (2006)—has also been included as a complementary statistical measure of uncertainty in the EVASPA daily ET estimates.”

**L315-319:**

“

$$SD_i = \left( \frac{\sum_{j=1}^n (Y_{i,j} - \mu_i)^2}{n} \right)^{0.5} ; CV_i = SD_i / \mu_i \quad (5. iii)$$

$$QCD_i = (Y_{Q3,i} - Y_{Q1,i}) / (Y_{Q3,i} + Y_{Q1,i}) \quad (5. iv)$$

”

“...  $\mu$  and  $n$  in the SD are the ensemble mean and ensemble size, respectively.  $Y_{Q1,i}$  and  $Y_{Q3,i}$  in the QCD are the respective first and third quartiles at  $i^{\text{th}}$  time step”

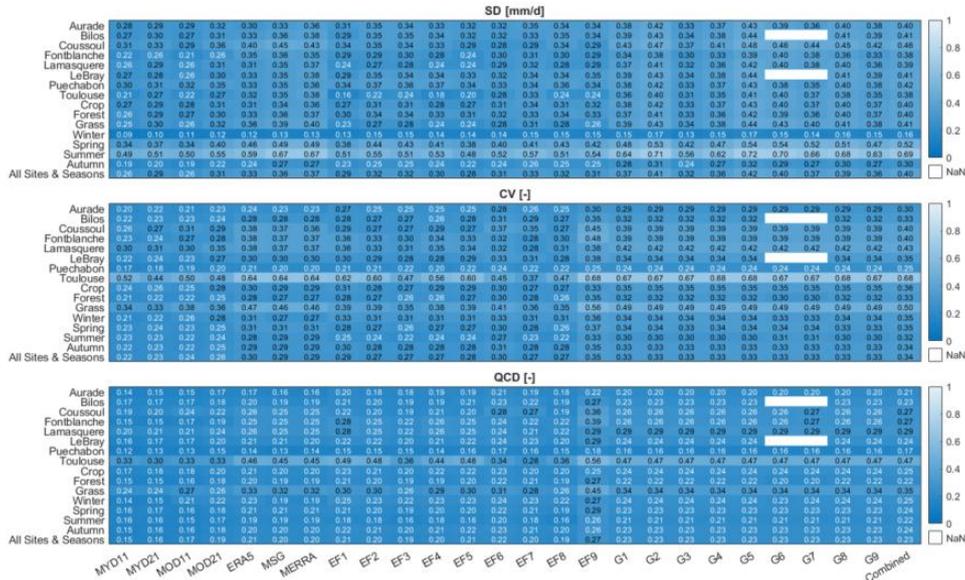
**L442-451:**

“The relative uncertainty (see CV and QCD) within the ensemble is also highest at this site. Conversely, the model also seems to overestimate the ET fluxes at the other grassland site (Coussoul). Although this site is relatively homogeneous in land cover (and land use), it is located in close vicinity to some water ponds, which may interfere with the thermal signals captured at the pixel scale. This likely introduces noise that propagates through the modelling process influencing the final estimates. As expected, seasonal performance of daily ET across all EVASPA member sets is consistently driven by the availability of radiation used for the surface turbulent fluxes, with the best performance (in RMSD and MAE) observed in winter. The performance then cyclically degrades over spring through summer and improves into autumn (Table 2, Figure 6). The absolute uncertainty (SD) in the estimated ensemble ETs shows a similar cyclical trend. The more informative relative uncertainty (as given by the relative SD or CV and the corresponding QCD) is however generally consistent throughout the seasons.”

**Table 2: Global performance of estimated daily evapotranspiration (average of all EVASPA ensemble members). Uncertainties within the ensemble are quantified using the median values of the standard deviation (SD), coefficient of variation (CV) and quartile coefficient of dispersion (QCD).**

	Interpolated daily ET estimates						non-interpolated daily ET estimates							
	RMSD [mm/d]	Willmott's D [-]	Bias [mm/d]	MAE [mm/d]	uncertainty (median)*			RMSD [mm/d]	Willmott's D [-]	Bias [mm/d]	MAE [mm/d]	uncertainty (median)*		
					SD [mm/d]	CV [-]	QCD [-]					SD [mm/d]	CV [-]	QCD [-]
Aurade	0.93	0.83	0.09	0.68	0.40	0.30	0.21	1.19	0.80	0.22	0.90	0.47	0.24	0.16
Bilos	0.84	0.86	-0.17	0.62	0.41	0.33	0.24	1.10	0.79	-0.08	0.84	0.50	0.28	0.20
Coussoul	0.97	0.63	0.58	0.72	0.46	0.40	0.27	1.15	0.58	0.71	0.90	0.53	0.37	0.26
Fontblanche	0.93	0.58	-0.19	0.74	0.38	0.40	0.27	1.07	0.52	-0.01	0.85	0.42	0.36	0.24
Lamasquiere	1.04	0.70	-0.41	0.70	0.40	0.43	0.30	1.27	0.62	-0.37	0.89	0.48	0.37	0.25
LeBray	0.74	0.85	-0.11	0.56	0.41	0.35	0.24	0.88	0.83	0.06	0.68	0.47	0.29	0.20
Puechabon	0.87	0.84	0.44	0.67	0.42	0.25	0.17	1.09	0.81	0.57	0.85	0.47	0.22	0.15
Toulouse	1.14	0.53	-0.75	0.82	0.38	0.69	0.48	1.35	0.38	-0.83	0.99	0.44	0.66	0.45
Croplands	0.99	0.78	-0.15	0.69	0.40	0.36	0.25	1.23	0.73	-0.07	0.89	0.47	0.29	0.20
Forests	0.87	0.80	-0.01	0.67	0.40	0.33	0.23	1.06	0.76	0.17	0.83	0.46	0.27	0.19
Grasslands	1.09	0.49	-0.32	0.79	0.41	0.51	0.35	1.27	0.40	-0.21	0.95	0.49	0.45	0.30
Winter	0.52	0.54	-0.14	0.39	0.16	0.35	0.25	0.60	0.49	-0.07	0.47	0.18	0.30	0.21
Spring	1.02	0.71	-0.23	0.79	0.52	0.35	0.24	1.25	0.63	-0.13	0.97	0.57	0.31	0.21
Summer	1.32	0.60	0.04	1.04	0.70	0.32	0.22	1.52	0.57	0.20	1.22	0.70	0.32	0.22
Autumn	0.70	0.66	-0.05	0.54	0.30	0.34	0.24	0.83	0.60	0.07	0.65	0.34	0.31	0.22
All sites and seasons combined	0.94	0.77	-0.09	0.69	0.40	0.34	0.23	1.16	0.73	0.03	0.87	0.47	0.31	0.22

\* Reported uncertainty are the median values over the uncertainty time series; e.g. for SD, the standard deviation is calculated for the ensemble at each time step, then the median of the SD time series is reported.



**Figure 6: Uncertainty heat maps (median of SD, CV, and QCD over the study period) for the different EVASPA models across all evaluation sites. More details on how the sets are constructed are contained in Figure S 1 in the supplements.**

See supplementary document for results including VIIRS based estimates

**Minor comments**

- No need to replicate the LST and NDVI acronyms, you already mentioned them at L54
- We have removed the unnecessary/repeated acronyms.
- This “relatively low” maybe would be better supported by some mention of relative uncertainty (standard deviation / average), to help the reader in contextualizing the potential error

Supporting text quantifying the (relative) uncertainty has been added. We have added relative measures of uncertainty in the text, i.e. the coefficient of variation  $CV = \text{standard deviation} / \text{average}$ , and the coefficient of quartile variation (also quartile coefficient of dispersion, QCD)

**L236-240:**

“We nonetheless acknowledge that SD (and the corresponding coefficient of variation, CV) can be misleading, especially when the analysed data exhibits non-symmetrical (specifically non-Gaussian) distribution. As such, the coefficient of quartile variation or quartile coefficient of dispersion (QCD)—which was recommended for non-normal distributions by Bonett (2006)—has also been included as a complementary statistical measure of uncertainty in the EVASPA daily ET estimates.”

**L315-319:** “

$$SD_i = \left( \frac{\sum_{j=1}^n (Y_{i,j} - \mu_i)^2}{n} \right)^{0.5} ; CV_i = SD_i / \mu_i \tag{5. iii}$$

$$QCD_i = (Y_{Q3,i} - Y_{Q1,i}) / (Y_{Q3,i} + Y_{Q1,i}) \tag{5. iv}$$

”

“...  $\mu$  and  $n$  in the SD are the ensemble mean and ensemble size, respectively.  $Y_{Q1,i}$  and  $Y_{Q3,i}$  in the QCD are the respective first and third quartiles at  $i^{\text{th}}$  time step”

- On a more general note, the manuscript is at times a bit too dense. Although this is understandable, given the scope of the analysis, this can hamper readability. Some sections, especially Section 4.2, feature long paragraphs with many interlinked concepts. Also some figures (e.g., 8, 10 or 11) feature either a complex caption, too many information or are hard to interpret.

Thank you for the feedback. The long sentences have now been shortened to improve readability. The captions have also been rephrased to make them clearer. Please refer to the revised manuscript with ‘tracked changes’.

For example,

“

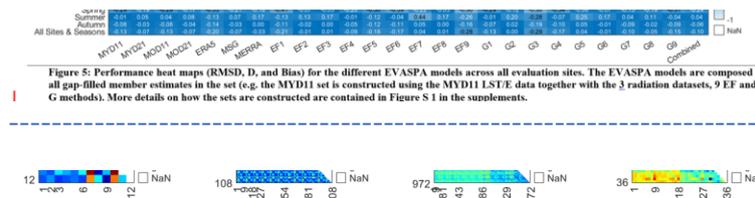


Figure 5: Performance heat maps (RMSD, D, and Bias) for the different EVASPA models across all evaluation sites. The EVASPA models are composed of all gap-filled member estimates in the set (e.g. the MYD11 set is constructed using the MYD11 LST/E data together with the 3 radiation datasets, 9 EF and 9 G methods). More details on how the sets are constructed are contained in Figure S 1 in the supplements.

\* E.g. for EF: 1-9 – all 9 EF methods using MYD11 LST, 10-18 – all EF methods using MYD21 LST, 19-27 – all EFs using MOD11 LST, and 28-36 – all EFs using MOD21 LST  
 \*\* I.e., net radiation - 3 × 4 combinations [radiation × LST products], G flux - 9 × 4 [G methods × radiation × LST products]; LE fluxes involve 9 × 9 × 3 × 4 combinations [G methods × EF methods × radiation × LST products]; EFs - 9 × 4 combinations [EF methods × LST products]

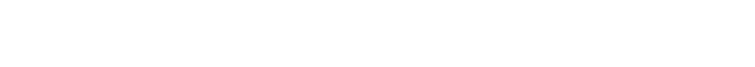
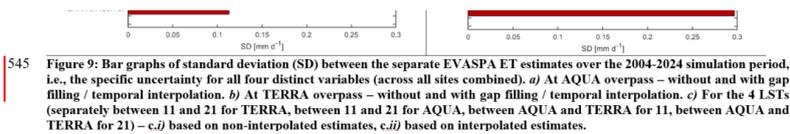


Figure 8: Heat maps of Willmott's D index [-] and MAE [ $W.m^{-2}$ ] (or MAE [-] for EF) between the EVASPA SEB members (net radiation - Rn, ground heat flux - G, latent heat flux - LE, and evaporative fraction - EF) over all sites and seasons combined. Heat maps of the metrics are calculated by taking each individual member vs all other ensemble members. \*Numbering in the heatmaps/ensemble is done according to the list of variables given in Table A 1. \*\*Distinct number of variable combinations in the heat maps are given in Table A 1.



545 Figure 9: Bar graphs of standard deviation (SD) between the separate EVASPA ET estimates over the 2004-2024 simulation period, i.e., the specific uncertainty for all four distinct variables (across all sites combined). *a)* At AQUA overpass – without and with gap filling / temporal interpolation. *b)* At TERRA overpass – without and with gap filling / temporal interpolation. *c)* For the 4 LSTs (separately between 11 and 21 for TERRA, between 11 and 21 for AQUA, between AQUA and TERRA for 11, between AQUA and TERRA for 21) – *c.i)* based on non-interpolated estimates, *c.ii)* based on interpolated estimates.

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