Response to Referee #1

Assessing evapotranspiration dynamics across central Europe in the context of land-atmosphere drivers

Comments from Reviewer 1

The manuscript titled "Assessing evapotranspiration dynamics across central Europe in the context of land-atmosphere drivers" evaluates evapotranspiration (ET) products derived from satellite remote sensing, modeling, and reanalysis data in conjunction with in-situ observations from Integrated Carbon Observation System (ICOS) stations in central Europe from 2017 to 2020. The study investigates the effects of varying land cover types, soil moisture (SM), and vapor pressure deficit (VPD) on ET dynamics, including the severe drought of 2018. It uses extended triple collocation methods to assess the accuracy of ET products, revealing notable differences among products under heterogeneous land cover conditions and during drought years. The findings highlight that ET variability is strongly influenced by VPD under nonlimiting soil moisture conditions and demonstrate that SEVIRI, ERA5-land, and GLEAM products show superior performance. The research provides insights into the suitability of various ET products for capturing land-atmosphere interactions and drought impacts across diverse land cover types. This manuscript presents valuable insights into evapotranspiration dynamics across central Europe, but significant revisions are needed before it can be considered for publication.

Thank you very much for all efforts and help to improve the manuscript. We tried our best to answer all comments satisfactorily.

Major comments:

 Limitations in the Performance Analysis of ET Products We discuss performance differences among ET products regarding varying climatic conditions, landcover types, and retrieval methods throughout the manuscript and especially in sec. 4. However, we added more information about the practical implications and objective of this study in sec. 4. and 5. of the manuscript.

The manuscript reveals significant performance differences among ET products, but it does not fully address how these discrepancies might affect the conclusions drawn, particularly under specific land cover types or climatic conditions. A deeper analysis of how these differences impact the study's practical implications is crucial.

In order to provide additional deeper analyses as suggested by the reviewer, we added the coefficient of variation for the binning plots for all ET products (see new figure 12) and added a discussion regarding the impact of the drought year on ET values in following sections.

 In-depth Analysis of Drought Year Effects

The analysis of the 2018 drought is valuable, but the study does not sufficiently explore how different ET products capture the effects of drought. A more detailed comparison of how products perform under drought conditions is needed, including which models and parameterizations are more sensitive to such extremes.

We discussed the impacts of the drought year on ET products in sec. 4.2. However, related to the comment above, we added statistics (CV) to the binning plots (Fig. 12) in order to provide deeper analyses of drought year effects.

 Improvement of Remote Sensing and Ground Observation Matching

There is an acknowledged mismatch between point-scale ICOS data and the coarser-resolution remote sensing products. The manuscript could benefit from a discussion on methods to address this issue, such as spatial downscaling or data fusion techniques, to improve the alignment of ground and remote sensing observations.

In the field of remote sensing, the spatial mismatch between satellite data and ground measurements is a well-known problem and somehow accepted assumption in order to be able to validate remote sensing products. In that sense, we agree on providing some discussion.

We added an explanation why we kept coarse-scale ET products at their original spatial resolution after line 186:

'The MODIS product with nominal spatial resolution of 500 m is aggregated to the 3 km footprint, while the SEVIRI, ERA5-land, GLDAS-2, and GLEAM products are maintained at their original spatial resolutions of 3 km, 9 km and 25 km, respectively. Although several downscaling methods and data fusion techniques exist for improving the spatial resolution of remote sensing products (Ha et al., 2013; Mahour et al., 2017; Peng et al., 2017), we decided to keep ET products with a spatial resolution lower than 3 km at their original resolution (i.e., GLEAM at 25 km). For one, the intention of this study is a comparison of well-known and established ET products and not an optimization of rescaled comparisons. Second, we did not want to include additional uncertainties potentially originating from the employed downscaling method or auxiliary datasets. Especially, downscaling approaches intend to statistically correlate coarsescale data and fine-scale auxiliaries, yielding to interpolation uncertainties and errors that cannot be quantified (Peng et al., 2017). All datasets are, however, temporally aggregated to daily time series in order to provide a temporal basis for comparison and analysis of the signal dynamics.'

Further, we added a discussion addressing this comment after line 511:

'One reason for the mismatch between the ICOS product and SEVIRI, GLEAM and ERA5-land is surely the spatial mismatch between the point-scale ground-based EC tower measurements and the remote sensing (3 km) or reanalysis (9 km) products. To capture vegetation stress, ecosystem health, and fine-scale variability in ET globally, adequate spatial (and temporal) resolutions are necessary. Here, detailed research regarding downscaling techniques, as reviewed in, e.g. (Mahour et al., 2017; Peng et al., 2017), that combine medium-scale ET data with fine-scale auxiliaries in order to improve the spatial resolution, are needed regarding its uncertainties and impact on product comparisons.'

Lastly, we draw a conclusion regarding this topic after line 647:

'In summary, when considering all conducted analyses together (spatial and temporal resolutions, product dependencies, ETC results, SM and VPD controls on ET), the remote sensing products SEVIRI and GLEAM as well as reanalysis product ERA5-land seems to provide most reasonable results compared to all other ET products, with SEVIRI providing a higher temporal and spatial resolution compared to GLEAM and ERA5-land. Hence, despite their coarse spatial resolution, GLEAM and ERA5-land are able to capture ET dynamics sufficiently even under drought conditions. Future research regarding data fusion techniques and downscaling approaches that combine coarse- or medium-scale ET data with fine-scale auxiliaries in order to improve the spatial resolution of certain ET products may help to decrease the spatial mismatch and optimize the comparison between point-scale field measurements and satellite remote sensing or modelling data.'

Even if we could largely extend the discussion on these methods, we prefer to limit it to the paragraphs written above since research regarding improved matching of remote sensing observations with ground measurements is out of the scope of this manuscript.

Considering extending the study period

The study period appears to be too short, and the comparison between satellite-based ET and local measurements seems insufficient. The authors might consider extending the study period and incorporating a comparison

We decided on purpose to analyze the years 2017 to 2020, for one, due to the availability of all datasets, and second, because we wanted to include detailed analyses regarding multiple ET products and varying climatic, landcover and site conditions. Further, we do not think that adding more years is necessary to draw conclusions on how well different ET products capture ET dynamics and drought events, as we already included varying climatic (one drought (2018) and one wet year (2017)) as well as landcover and site conditions (eight stations with different landcover and climate conditions as well as levels of landcover heterogeneity).

	across different timeframes. This would enable the use of the water balance principle for a more comprehensive evaluation.	Analyzes across different time frames and evaluation regarding the water balance principle is out of the scope of the manuscript, which intends to compare well-known and established ET products that are commonly used in climatic research studies.
Minor comments 1. At lines 34-35: please add expla-		The explanation is found in the following sentence. We rephrased the sentences to provide the explanation right away:
	nation (why).	'The greatest deviations are found at the agricultural-managed sites Selhausen (Germany) and Bilos (France), with the former also showing the highest potential dependencies (error cross-correlation) between the ET products. Hence, o ur results indicate that ET products differ most at stations with spatio-temporal varying land cover conditions (varying crops over growing periods and between seasons).'
2.	At line 34: please add the specific	Of course, we added the specific number:
	number to describe "the highest potential dependencies (error cross-correlation)".	'The greatest deviations are found at the agricultural-managed sites Selhausen (Germany) and Bilos (France), with the former also showing the highest potential dependencies (error cross-correlation (ECC)) between the ET products (up to 7.6 and outside the acceptable range of -0.5 < ECC < 0.5).'
3.	At lines 36-38, please explain why.	The explanation is found in the following sentence. Due to the more complex heterogeneity of land covers at agricultural stations, the estimation of ET is more complex and challenging, while needle-leaved stations have less complexity due to rather stable vegetation conditions. We rephrased the sentences to be more precise:
		'This is since complex heterogeneity complicates the estimation of ET, while ET products agree well at evergreen needle-leaved stations with less temporal changes throughout the year and between years.'
4.	Please add the longitude and latitude in Figure 1.	Done. We added coordinates to the map in figure 1.
5.	Please add a table that describes all the sites, including their longitude, latitude, land cover type, altitude, and other relevant details.	Done. We added a table S1 in the supplement providing the coordinates, ecosystem type, altitude, climate zone and mean annual precipitation and air temperature for all stations. Details on percentages of different land cover types for every station are already given in supplement table S2 (formerly tab. S1).
6.	At line 150: Please explain why you used the standardized pre-	Done. We added some sentences to explain why we used the SPEI instead of SPI or other indices after line 159:
	cipitation-evapotranspiration index (SPEI).	'The choice of SPEI to identify drought conditions instead of the standardized precipitation index (SPI) or other indices (i.e., Palmer drought severity index) is due to the fact that the SPEI considers implicitly temporal changes in ET and hence, temperature, which is relevant for identifying abnormal (drought) conditions and for this study with focus on ET variations. Previous studies showed that not only the lack of precipitation defines drought events but also the

		level of temperature and consumption of rainfall by evaporation and/or transpiration (Vicente-Serrano et al., 2010).'
7.	If the time step is from hourly to 8-day, consider generated them at a shorter time step rather than monthly which loose too much information.	We agree that analyzing monthly time steps would be temporally too coarse. That is why in the manuscript, we are analyzing daily and 8-daily time steps, never monthly.
8.	At lines 206-225: The spatial resolution differences among these ET products are quite significant.	This comment is related to comments #3 and #4 above, where we provide our changes implemented in the manuscript to account for them.
	The authors should consider using methods to standardize all the ET products to a common resolution or use products with longer time periods. Otherwise, direct comparisons may not be valid or meaningful.	We agree that the change across spatial resolutions between different products are quite significant and we tried to account for it to certain degree as we upscaled high-scale ET products (e.g., MODIS at 500 m) to the baseline resolution of 3 km. And we decided to keep coarsescale ET products (> 3 km, e.g., GLEAM at 25 km) at their original resolutions instead of rescaling, since, for one, the downscaling from 25 km to 3 km would be a very big step and we did not want to include additional uncertainties from the employed rescaling method, and second, end users who would use only one ET product within their analyses alongside other variables would not rescale the ET product to one specific resolution and hence, loose the conclusions drawn from our analyses if we would compare all products at one rescaled resolution.
		We think the comparison is meaningful and valid as we compare well-known and employed ET products under varying climatic conditions and for multiple landcover classes at several stations all across Central Europe. And we were able to show that despite the coarse spatial resolution of GLEAM, the product is able to capture ET dynamics quite well compared to other products, which is an important conclusion for the scientific community and especially for studies that are based on the GLEAM ET product only.
9.	Table 1: please change as three lines table format and added other dataset's information such as soil moisture and SPET etc.	Done, we included other products (VPD, SM, SPEI) in table 1. However, we did not change the format of the table as we like it this way.
10.	. At lines: 228-233 The removal of the seasonal signal may not be necessary, as the study period is too short for such adjustments.	We understand this point and agree that for longer time series, detrending has much more impact. We would like to keep the anomalies in order to compare detrended ET time series, e.g., figure 8 & 9, even if the time series considered is rather short compared to real climatological analyses.
11.	. At lines 295-296: Please provide the specific values for the highest R ² and lowest RMSE and lowest percentage bias, PBIAS here.	We provide the detailed numbers for R ² , RMSE and PBIAS in the following sentences. Since every station provides multiple statistics between all products (as shown in supplement figures S5-S7), we first wanted to highlight specific stations with outstanding statistics and then provide more details with actual numbers for the products.

	Otherwise, the sentences would get very long and complex, hence, we thought it might be easier for the reader to follow.
12. Figure 4. Please added some statices numbers in every panel.	Figure 4 shows the time series of all ET products and for all stations. Meaning, we are showing one single time series of every ET product and for every station individually. We are not sure what kind of statistics would make sense for single time series that would provide any useful information.
13. Same as Figure 4	We guess this comment is related to figure 5 and that we should add statistics in figure 5, same as in figure 4? Or to which figures is this comment related to?
	The statistics (correlation, error & percentage bias) for figure 5 as well as similar supplement figures S2-S4 are given in supplement figures S5-S7. Since the figures give already a lot of information, we provide the statistics in separate figures.

Response to Referee #2

Assessing evapotranspiration dynamics across central Europe in the context of land-atmosphere drivers

Comments from Reviewer 2				
Review Report for Manuscript ID: egusphere-2024-3386	Thank you very much for reviewing our study and for outlining the relevance of the manuscript. We tried our best to provide satisfactorily answers to all comments in order			
Title: Assessing Evapotranspiration Dynamics Across Central Europe in the Context of Land-Atmosphere Drivers	to improve and strengthen the manuscript.			
General Comments				
This study provides a comprehensive evaluation of evapotranspiration (ET) products across central Europe using a combination of in-situ, remote sensing, and reanalysis datasets. The authors analyzed the performance of multiple ET datasets under different climatic conditions, particularly focusing on the severe drought. The study effectively addresses a research gap by assessing the agreement and discrepancies between ET products in the context of soil moisture (SM) and vapor pressure deficit (VPD)				

interactions. The manuscript is relevant for researchers studying land-atmosphere interactions, hydrology, and ecosystem responses to climate extremes. However, several key areas require further clarification and improvement to strengthen the manuscript before publication. The authors should clarify the rationale for dataset selection, improve the discussion on physical interpretability, and provide additional insights into the role of vegetation stress and uncertainty quantification.

Major Comments

Justification for Selected ET Products, the authors compare ET estimates from various remote sensing and modeling products (MODIS, SEVIRI, GLEAM, ERA5-land, GLDAS). However, it would be beneficial to explicitly justify the selection of these specific products over other alternatives such as FLUXCOM, ETMonitor, EB-ET. MODIS is kind of ET more relying on optical data. GLEAM is based on microwave data. Please check other thermal ET product. One could be EB ET. Chen et al. 2021, Remote sensing of global daily evapotranspiration based on a surface energy balance method and reanalysis data. Journal of Geophysical Research: Atmospheres, 126(16): e2020JD032873.

Thank you for naming several other ET products and providing the reference for the EB-ET product.

The reviewer mentioned products are higher order ET products derivated from either other ET products or satellite products (e.g., Fluxnet ET observations, MODIS), and often provide data outside the time period analyzed in this study. In this study, we compare ET products directly retrieved from the observations and mostly from one single sensor (e.g., ICOS LE observations, MODIS optical data, ERA5-land reanalysis, SEVIRI) between 2017-2020. However, we can add the proposed products to our list, when we introduce existing ET products, and further discuss their advantages and disadvantages.

We added sentences to justify in more detail why we choose these ET products (most commonly known and employed ones) and acknowledge the existence of other products, e.g.,:

Lines 105-106: 'In this study, we first compare the most common ET products from field measurements, modelling, and remote sensing across 104 central Europe for the period 2017 to 2020. These selected six products (ICOS, MODIS, SEVIRI, ERA5-land, GLDAS-2, and GLEAM) are well-known, commonly employed, and freely available.'

Line 532: 'Although there exist other ET products from remote sensing and modelling, e.g., (Jiménez et al., 2011; Mueller et al., 2013; Fisher et al., 2020; De Santis et al., 2022; Yu et al., 2023), the examined ET products in this study are appropriate when addressing global analyses since other products have either a more coarse spatial or temporal resolution (Yu et al., 2023), are limited to clear sky conditions (De Santis et al., 2022), which prohibits continuous time series of ET measurements, or are higher order derivates from either field measured or merged remote sensing based products (Jung et al., 2019; Chen et al., 2021).'

Additionally, the manuscript should discuss the potential biases associated with the retrieval algorithms used in each dataset and how these may affect ET estimates under different climatic conditions.

Certainly, we added a paragraph for discussing on potential biases associated with the retrieval algorithms for every ET product in section 2.2.

Physical Interpretability and Model Dependencies, the study provides robust statistical comparisons but lacks a deeper discussion on the physical implications of the observed differences. For example, why do some products perform better at evergreen needle-leaved sites compared to agricultural sites? How do land cover heterogeneity and seasonal changes influence model uncertainties?

We discuss the performance of the different ET products for varying landcover classes throughout the manuscript, and discuss the influence of land cover heterogeneity on retrieval results in section 4. of the manuscript. More thorough analyses and discussion regarding the physical interpretability and model dependencies with that many ET products would lengthen the paper significantly, which is impractical and potentially more interesting when focusing on two or three ET products.

Since GLEAM incorporates reanalysis and satellitebased observations, its correlation with other datasets like ERA5-land and GLDAS-2 might be inflated. Have the authors accounted for interdependencies between products in their error analysis? We accounted for the interdependencies between products and calculated the error cross-correlation (ECC) between all products (see sec. 2.3.1.) in order to statistically validate the interdependencies. We discuss this in sec. 3.1. as well as 4.1., and give the ECC results in supplement figure S8.

The role of vegetation stress and physiological controls (e.g., stomatal closure) in driving ET reductions during drought should be better discussed, perhaps using additional to support this point.

Certainly, we added more discussion on the role of vegetation stress and physiological controls in ET reductions during the drought year 2018 throughout the manuscript, e.g., section 4.2. & 5.

Evaluation of Uncertainty and Error Cross-Correlation (ECC), The extended triple collocation (ETC) analysis is a valuable approach, but some ECC values are quite high, particularly at agricultural sites. The manuscript should explicitly discuss how ECC influences the reliability of the results and whether certain datasets may be inherently dependent.

We discussed the ECC results and potential interdependencies between ET products in sec. 3.1. as well as 4.1., and showed that ETC results are uncorrelated to ECC results, since at station DE-Rus, which gave high ECC results (potential strong interdependencies), ETC results do not reflect any interdependencies.

Clarity of Figures and Statistical Significance, the scatter plots and time series comparisons are informative, but additional clarity is needed in figures showing product inter-comparisons (e.g., Figures 4, 5, 6). Including a statistical significance test for differences between ET products would enhance the rigor of the results.

We included significance tests in figures 8 and 9 since we agree that some analyses need statistical significance information to discuss the results reliably.

However, figure 4 shows the time series of all ET products and for all stations. Meaning, we are showing one single time series of every ET product and for every station individually. We are not sure what kind of statistics would make sense for single time series that would provide any useful information.

In figure 5, we show the scatterplots, for which we give the Pearson's correlation coefficient (R²), root-mean square error (RMSE) and percentage bias (PBIAS) between each

which we discuss in sec. 3. and 4. Well taken, we tried to improve the grammar and style. **Minor Comments** Grammar and Style: Some sentences are long and complex, making them difficult to follow. Consider simplifying and improving readability. For example: The ICOS network has undertaken a large effort to ensure high-quality LE measurements, which are comparable among different ICOS stations.". Suggested revision: "The ICOS network has made significant efforts to ensure consistent high-quality LE measurements across stations.". Line 555 Grammar mistake, This is, products were most consistent with each other at stations with less complex land cover conditions and changes throughout the seasons (the evergreen needle-leaved stations DE-Ruw and FI-Let). Line 49, rephrase the sentence 'Since precipitation Done, we rephrased the sentence: (P) 'and evaporation are the two key components of 'Evapotranspiration (ET) is an important proxy for analyzthe global water cycle' (Miralles et al., 2011), aning water stress and its effects on ecosystems since precipother important proxy for analyzing water stress and itation (P) 'and evaporation are the two key components its effects on ecosystems is evapotranspiration (ET).' of the global water cycle' (Miralles et al., 2011).' Line 80, optical, thermal, infrared, or microwave ob-Thank you for pointing this out. We included two addiservations are used to derive ET based on surface tional references, explicitly addressing thermal based ET energy balance, physical and empirical models and the surface energy balance method: (Bayat et al., 2021, 2024; Rahmati et al., 2020; Zhang 'Although it is not directly measurable from remote senset al., 2016). The cited reference does not include ing acquisitions, optical, thermal, infrared, or microwave thermal observation based ET from surface energy observations are used to derive ET based on surface enbalance method. ergy balance, physical and empirical models (Zhang et al., 2016; Rahmati et al., 2020; Singh et al., 2020; Bayat et al., 2021; Bhattacharya et al., 2022; Bayat et al., 2024).' These references are: Singh, R.P., Paramanik, S., Bhattacharya, B.K. et al. Modelling of evapotranspiration using land surface energy balance and thermal infrared remote sensing. Trop Ecol 61, https://doi.org/10.1007/s42965-020-42-50 (2020).00076-8 Bhattacharya, B. K., Mallick, K., Desai, D., Bhat, G. S., Morrison, R., Clevery, J. R., Woodgate, W., Beringer, J., Cawse-Nicholson, K., Ma, S., Verfaillie, J., and Baldocchi, D.: A coupled ground heat flux-surface energy balance model of evaporation using thermal remote sensing observations,

product and for each station in supplement figures S5-S7,

	Biogeosciences, 19, 5521–5551, https://doi.org/10.5194/bg-19-5521-2022 , 2022.
Terminology Consistency: The terms "ET estimation," "ET retrieval," and "ET modeling" are used interchangeably. It would be beneficial to define them more precisely and use consistent terminology throughout the manuscript.	Done. We used only one terminology (ET estimation) throughout the manuscript. The term 'modelling' is only used for modelled ET products (e.g., GLDAS-2). The term 'retrieval' is only used when talking about the retrieval algorithm/method of the ET products.
Temporal Aggregation Effects: Some ET product has a lower temporal resolution than other datasets. Have the authors checked whether this affects the observed discrepancies, if upscaled to 15 days, even monthly temporal resolution?	Yes, we have checked the effect of temporal aggregation. In the manuscript, we included some of these analyses in figure 9 and supplement figure S10 by comparing the Kernel density estimates of ET anomalies on daily and 8-daily time scales. Here, we clearly see a difference in the density curves and hence, included it in the publication as these are the main temporal scales included in the paper. As most products provide high temporal resolution (< daily) except for MODIS, which provides 8-daily ET data, analyzing other temporal scales (15-daily, monthly) is out of the scope of the current manuscript.
Line 558, The authors wrote that: The remote sensing products, SEVIRI, MODIS, and GLEAM, performed equivalently well or even better than the in-situ measured (ICOS), I don't understand why the remote sensing products can be better than in-situ measured data? This is confusing readers. How can a satellite ET product be better than measurement?	Thank you for pointing this out. We are not saying that remote sensing products are better than in-situ observations. What we mean is, that for our specific study design (3 km footprint, daily analyses), the performance of the remote sensing products is overall more comparable and consistent among all investigated datasets, including ICOS. We discussed the reasons for this, e.g., in lines 513-518, also providing the drawbacks of in-situ measured ET observations at ICOS stations. We can rephrase the sentence in lines 621-623 in order to clarify the meaning.
	'The remote sensing products, SEVIRI, MODIS, and GLEAM, performed equivalently well or even better than the insitu measured (ICOS), modelled (GLDAS-2) or reanalysis (ERA5-land) products for this specific study concept (3 km footprint, daily analyses).'
Line 514, ET is more controlled by atmospheric demand rather than atmospheric supply, I can understand when the atmosphere is warming, it will need more vapor evaporated from ground. This could be a kind of atmospheric demand, but do not understand what is atmospheric supply? What kind of supply from atmosphere can influence ET? Are you saying precipitation? Please rephrase this sentence to make it clear.	Thank you for pointing this out. We actually meant soil water supply (soil moisture) as we see no dependency of ET on SM but variations of ET with VPD during wet years. We can rephrase the sentence accordingly: 'Here, our results indicate that ET is more controlled by atmospheric demand rather than water supply from atmosphere (precipitation) and soil (soil moisture) as reported also by Zhou et al., (2019).'

Line 523, Further, results show that VPD and SM are	Thank you, we included De Santis et al., 2022 as additional
negatively coupled during extreme events as reported also by (Zhou et al., 2019)-à by Zhou et al.	reference.
2019. Same as reported by (De Santis et al., 2022).	
2019. Same as reported by (De Santis et al., 2022).	
Section 4, there are many other global ET product,	As mentioned in our second answer above, we mentioned
which are not discussed. Please check and compare	other ET products and included additional information
them.	why we choose to compare these ET products.