## Revision Notes, egusphere-2025-3146

Dear Armin,

We would like to express our sincere gratitude for your time and feedback on our manuscript, " Rapid Flood Mapping from Aerial Imagery Using Fine-Tuned SAM and ResNet-Backboned U-Net."

## The author's reply to the comment:

Comment	Response
Dear Hadi,	Dear Armin,
Thank you—please don't get me	Thank you once again for your
wrong—you are one of our remote	valuable feedback. We truly
sensing community, and your work is	appreciate the time and effort you have
good. I don't want to dwell on the	taken to engage with our work.
similarity aspect, as we've already	In our work, we not only compared U-
explored comparisons between SAM and	NET and SAM, but also evaluated two
UNet50 (with ResNet backbone) in the	types of input prompts in SAM (points
context of water segmentation using	and bounding boxes) and two types of
close-range remote sensing images (from	backbones for U-NET (ResNet-50 and
UAVs, smartphones, and handheld	ResNet-100). We agree that including
cameras, within a 1–300 meter range)	a direct comparison with your findings
(https://doi.org/10.1109/ACCESS.2024.	will make our paper more
3385425). However, I'd like to suggest	comprehensive, and we will add this
that you consider referencing our results,	comparison to the results section of
as they align with your findings and could	our revised manuscript.
strengthen your discussion section. For	Regarding the ViT backbone, we also
instance, we observed that SAM performs	evaluated multiple variants and
very well in general. However, when	observed that their effectiveness for
segmenting images from the same area,	flood-affected area segmentation was
UNet actually produced even better	comparable. To optimize
results. This nuance might enrich your	computational resources, we selected
discussion, especially when highlighting	ViT-Base, as it provides a favorable
the practical performance differences	balance between accuracy and
between models.	efficiency.

Another point is related to the computational aspects: SAM typically operates on 1024×1024 patches, and when fine-tuning with a frozen ViT backbone, it still requires significant computational resources. The choice of ViT backbone also matters—ViT-H is quite heavy and not ideal for fine-tuning, whereas the smaller variants (like Tiny ViT and Medium ViT) tend to perform better with fewer resources.

Lastly, regarding datasets: One could argue that with SAM, we might not need large annotated datasets anymore. While SAM reduces the need for manual annotation, I would still say that datasets are necessary. The real question is: how many do we actually need? To help address this, you might consider referencing this recent paper by Professor Anette Eltner from Dresden University: https://doi.org/10.1080/01431161.2025.2 457131

It discusses the sensitivity of model performance to dataset size and could help you frame this as a potential advantage of SAM over UNet.

I'd love to hear your thoughts and see how you might incorporate some of these ideas into your discussion.

Warm regards,

Armin

We appreciate also your recommendation of Professor Anette Eltner's recent article. It provides an excellent perspective on dataset size requirements. The number of labeled images needed depends on complexity of the task. For instance, in our previous study on erosion and deposition segmentation (https://doi.org/10.1016/j.catena.2025 .108954) we worked with about 400 labeled images and observed clear performance gains with increasing dataset size, an effect that was particularly relevant given the higher complexity of that task compared to This flood mapping. increased complexity was because eroded and non-eroded soil often have very similar visual characteristics, whereas flooded areas are usually more distinct from their surroundings.

Thank you again for your thoughtful suggestions. They will certainly help us improve the clarity and impact of our paper.