Appendix

Fig. 1 gives some examples of segmentation results by VM [1], Sli2Vol [7], Vol2Flow [2], and our method on on the Decath-Liver, Decath-Spleen, Decath-Heart, and Decath-Brain Tumours datasets [5], with GT given as reference. From the visual segmentation results, we can observe that the segmentation results produced by our method are significantly better than those produced by the known mask propagation methods on all the four datasets. In particular, our method generates accurate segmentation results, while the other mask propagation methods generate false negatives on the Decath-Liver and Decath-Brain Tumours datasets. This demonstrates the effectiveness of our method.

Fig. 2 presents some visual results from various segmentation foundation models (i.e., SAM [3] and MedSAM [4]) or interactive segmentation tools (i.e., ScribblePrompt [6]) on the Decath-Liver, Decath-Spleen, and Decath-Heart datasets [5]. From the visual segmentation results, we can observe that the segmentation results produced by ScribblePrompt [6] are better than those produced by the known segmentation foundation models on all the three datasets. Fig. 3 shows some examples of segmentation results generated by our method when combining different key components on the Decath-Liver and Decath-Spleen datasets [5]. From the visual segmentation results, we can observe that the segmentation results are improved when different key components are combined. This demonstrates the effectiveness of the proposed components.

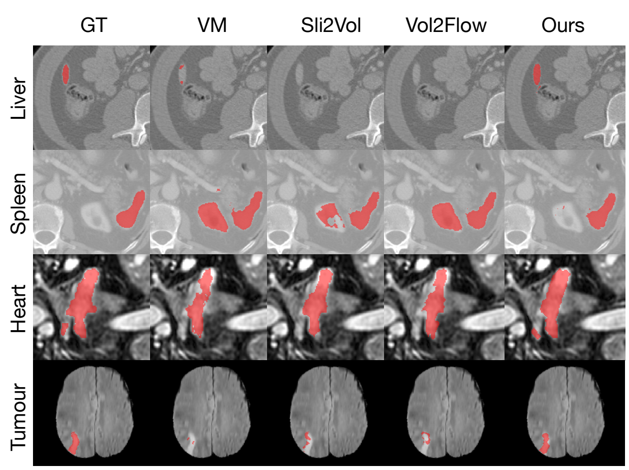

Figure 1. Examples of segmentation results by different methods on the Decath-Liver, Decath-Spleen, Decath-Heart, and Decath-Brain Tumours datasets [5].

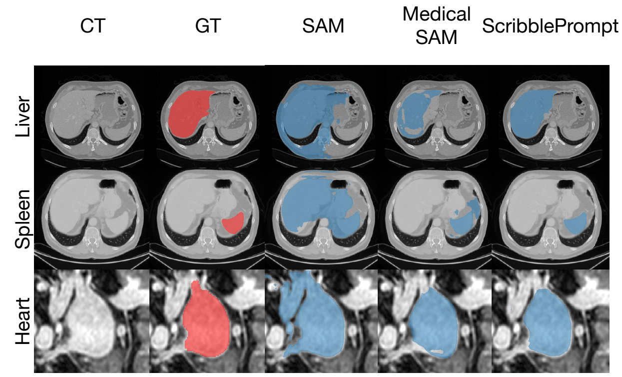

Figure 2. Examples of segmentation results by various segmentation foundation models or interactive segmentation tools on the Decath-Liver, Decath-Spleen, and Decath-Heart datasets [5].

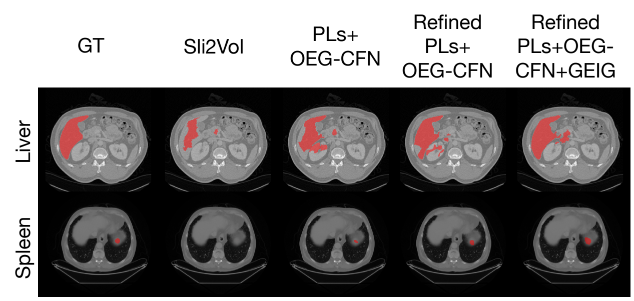

Figure 3. Examples of segmentation results generated by our method when combining different key components on the Decath-Liver and Decath-Spleen datasets [5].

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