



RGB-D Video Object Segmentation via Enhanced Multi-store Feature Memory

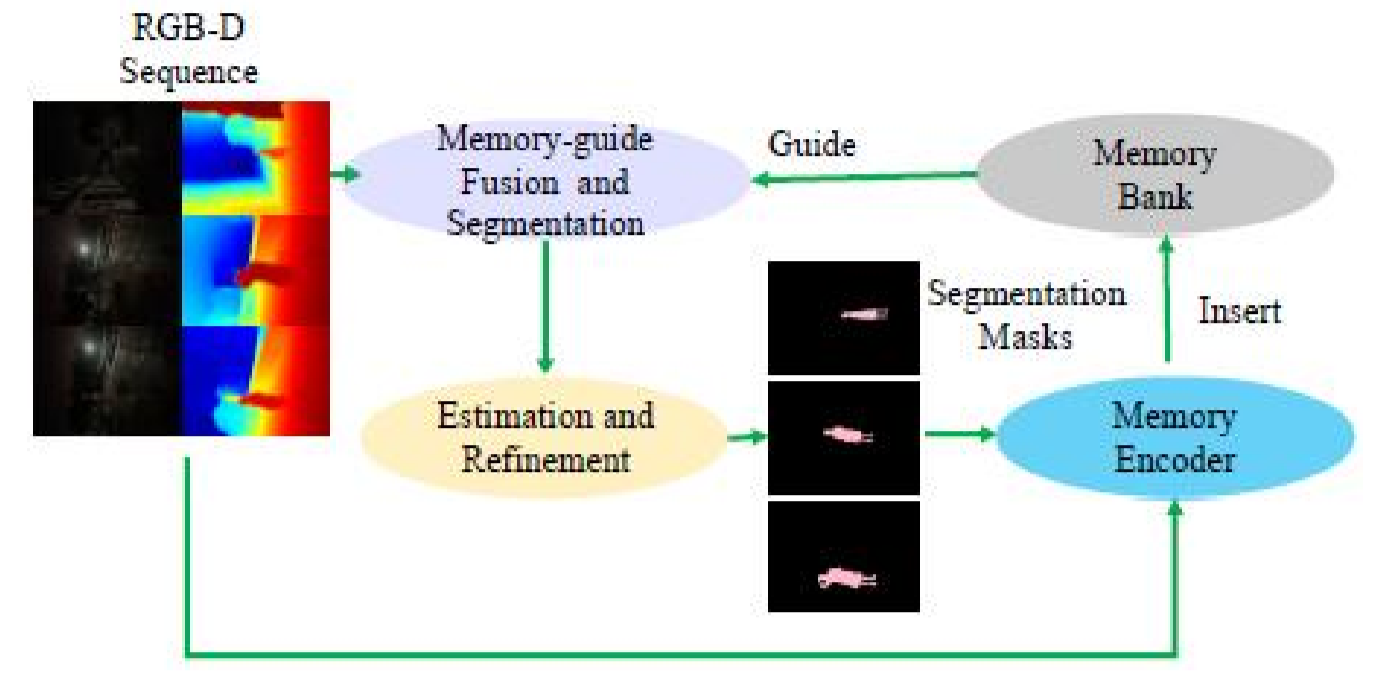
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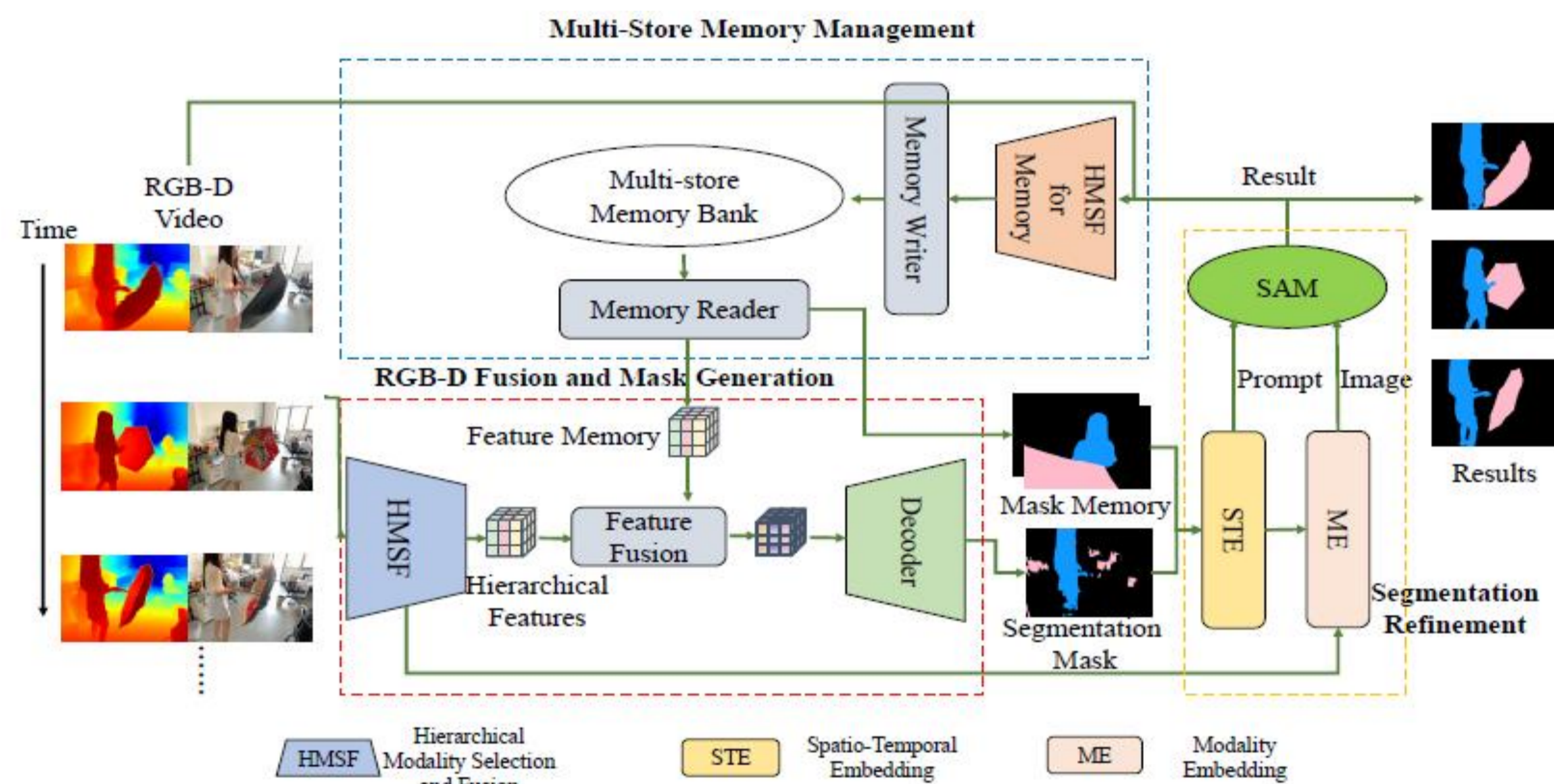
Introduction

RGB-Depth (RGB-D) Video Object Segmentation (VOS) combines RGB and depth data. RGB offers visual details like color and texture, while depth provides the distance from the camera to the object. We propose a RGB-D VOS method **via multi-store feature memory** for robust segmentation. Specifically, we design the **hierarchical modality selection and fusion** and a **segmentation refinement** module, significantly improving the robustness of segmentation.



Method

The proposed method comprises three modules: **RGB-D fusion and mask generation module**, **segmentation refinement module**, and **multi-store memory management module**. Specifically, the RGB-D fusion and mask generation module aims to fuse RGB-D dual-modality features and integrate them with multi-store feature memory to produce segmentation results. The segmentation refinement module flexibly utilizes the SAM to refine segmentation results and ensure more accurate results as memory to guide subsequent segmentation. The multi-store memory management module encodes and stores both RGB-D images and segmentation results as feature memory.



Experiments

Dataset: ARKitTrack

Metrics: Region similarity (J), Contour accuracy (F), J&F

Comparison with the SOTA: The method is superior to all the RGB-D VOS methods

Ablation Study: The experiment demonstrates the effectiveness of each component of the method

Ablation	STE	ME	$\mathcal{J}_M \uparrow$	$\mathcal{F}_M \uparrow$	$\mathcal{J}\&\mathcal{F} \uparrow$
			0.617	0.680	0.649
✓			0.637	0.691	0.664
✓	✓		0.651	0.702	0.677
✓	✓	✓	0.673	0.723	0.698

Comparison	Methods	Year	$\mathcal{J}_M \uparrow$	$\mathcal{F}_M \uparrow$	$\mathcal{J}\&\mathcal{F} \uparrow$
	STCN [5]	2021	0.489	0.560	0.525
	AOT [41]	2021	0.555	0.627	0.582
	RPCM [38]	2022	0.492	0.527	0.509
	QDMN [21]	2022	0.276	0.337	0.306
	XMem [4]	2022	0.541	0.565	0.553
	STCN_RGBD [5]	2021	0.498	0.574	0.537
	XMem_RGBD [4]	2022	0.617	0.680	0.649
	SAMTrack_RGBD [7]	2023	0.445	0.463	0.454
	ARKitTrack [46]	2023	0.625	0.698	0.662
	Ours	-	0.673	0.723	0.698

