

Strike the Balance: On-the-Fly Uncertainty based User Interactions for Long-Term Video Object Segmentation

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Scan for Code

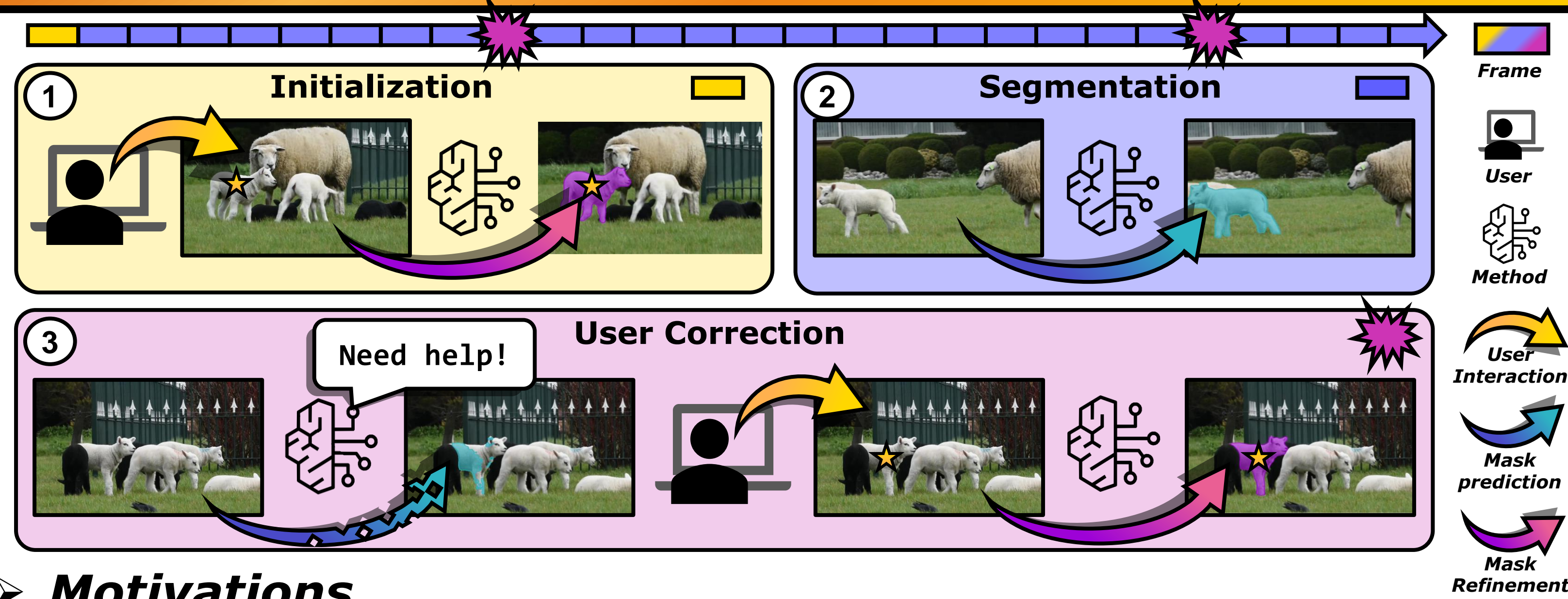


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ACCv
2024

TL;DR



Motivations

- Maximize tracking for long-term VOS
- Minimize human oversight (only at delicate events)
- Allow user corrections on-the-fly with user-clicks

Contributions: **ziVOS** and **Lazy-XMem**

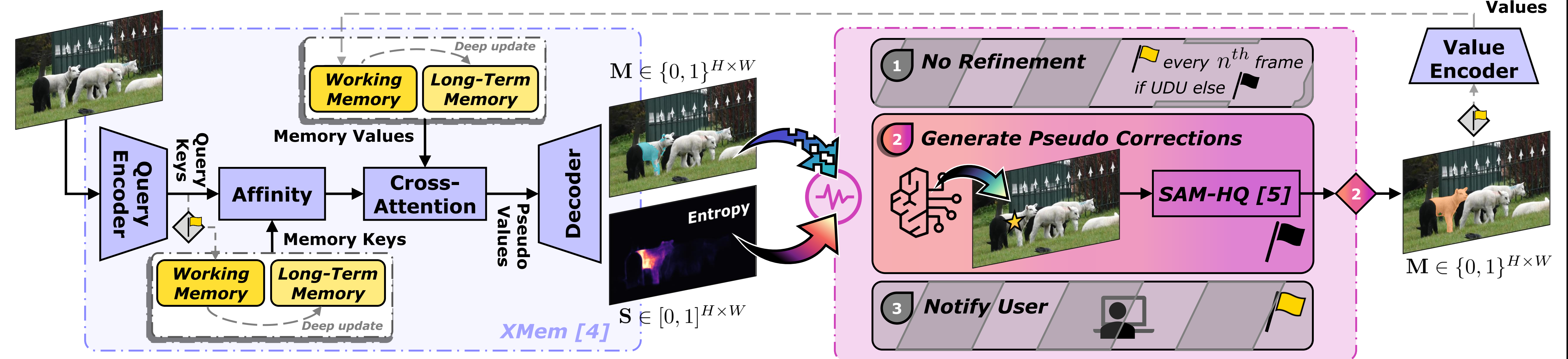
- New Task: Lazy Video Object Segmentation
- On-the-fly assessment of tracking accuracy
- Generate pseudo- and support user-corrections on-the-fly

Results

- Increase accuracy (by 11% $\mathcal{J}\&\mathcal{F}$) and robustness (by 10% $R@0.1$)
- Interact with only 1% of the dataset to improve robustness

Methodology – Lazy-XMem

Overview



1. Uncertainty Estimation

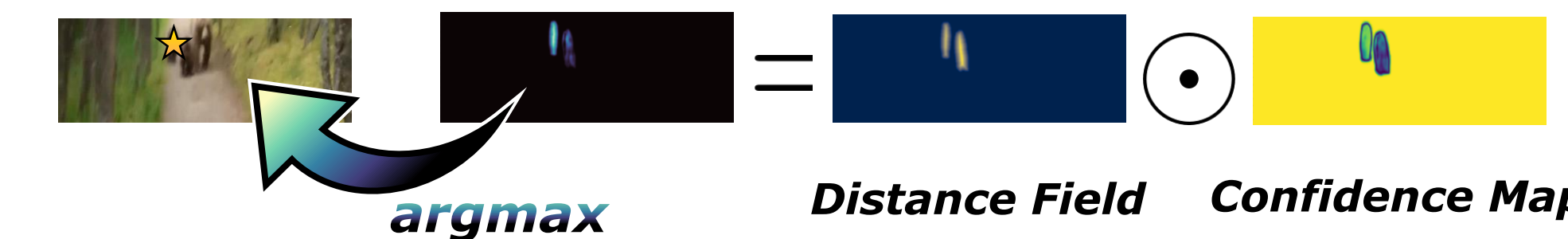
- Pixel-level uncertainty [1]



- IoU vs Uncertainty (Correlation? Yes!)

2. Mask Refinement

- Sam-HQ [5]
- Issue corrections
- Pseudo-corrections on-the-fly



3. Memory Update

- Uncertainty driven update (UDU)
- Interaction driven update (IDU)

Qualitative Results

Pseudo-Corrections

✔ Successes

✘ Failures



User-Corrections

✔ Successes

✘ Failures



Metrics

Performance

- Accuracy ($\mathcal{J}\&\mathcal{F}$)
- Robustness ($R@_{\tau_{IoU}}$)

User-Workload

- Number of corrections (NoC)
- Interaction density index (IDI)
- Average correction interval (ACI)

$$R@_{\tau_{IoU}} = \frac{1}{|\mathcal{O}|} \sum_{o \in \mathcal{O}} \frac{1}{|\mathcal{F}_o|} \sum_{f \in \mathcal{F}_o} 1_{[IoU(M_f^o, GT_f^o) \geq \tau_{IoU}]} \quad ACI = \sum_{o \in \mathcal{O}} \frac{1}{|\mathcal{F}_o|} \sum_{i=1}^{|\mathcal{F}_o|} \sum_{j=1}^i n_j$$

Quantitative Results – ziVOS

Benchmark

Method	$\mathcal{J}\&\mathcal{F}$	Robustness			User-Workload		
		$R@0.1$	$R@0.25$	$R@0.5$	ACI	NoC	IDI
<i>sVOS Methods</i>							
QDMN [6] (ECCV 2022)	44.2	47.8	45.5	36.2	-	-	-
XMem [4] (ECCV 2022)	52.8	57.0	55.0	49.0	-	-	-
DEVA [7] (ICCV 2023)	55.1	63.6	59.3	52.4	-	-	-
Cutie-base [8] (CVPR 2024)	57.0	59.2	57.8	52.4	-	-	-
Cutie-small [8] (CVPR 2024)	57.6	58.6	57.0	52.5	-	-	-
Lazy-XMem [†] (ours)	56.4	58.8	56.8	50.6	-	-	-
<i>ziVOS Methods</i>							
Rand-Lazy-XMem	61.3	67.9	65.8	59.3	5.17	335	17.9
Lazy-QDMN	52.7	58.2	52.0	42.9	5.64	360	16.7
Lazy-XMem (ours)	64.3	70.2	67.8	62.3	5.02	325	18.4

[†] No User Corrections

Ablations

Configuration												
	<i>Pseudo</i>		<i>User</i>			Robustness			User-Workload			
UDU	Corr.	IDU	Corr.	IDU	$\mathcal{J}\&\mathcal{F}$	$R@0.1$	$R@0.25$	$R@0.5$	ACI	NoC	IDI	
-	-	-	-	-	52.8	57.0	55.0	49.0	-	-	-	
✓	-	-	-	-	54.7	56.3	54.5	50.0	-	-	-	
✓	✓	-	-	-	56.4	58.8	56.8	50.6	-	-	-	
✓	✓	✓	-	-	53.1	57.0	55.1	49.6	-	-	-	
✓	-	-	✓	-	55.6	58.2	56.4	51.8	7.80	507	12.6	
✓	-	-	✓	✓	62.9	67.8	66.2	60.9	5.05	327	18.3	
✓	✓	-	✓	✓	64.3	70.2	67.8	62.3	5.02	325	18.4	
✓	✓	✓	✓	✓	64.3	70.1	68.2	62.1	5.91	352	17.3	

Results from LVOS [3]

[1] Shannon, C.E. A mathematical theory of communication. The Bell system technical journal, 1948.

[2] Pont-Tuset, J. et al. The 2017 DAVIS challenge on video object segmentation. arXiv, 2017.

[3] Hong, L. et al. LVOS: A benchmark for long-term video object segmentation. ICCV, 2023.

[4] Cheng, H.K., Schwing, A.G. XMem: Long-term video object segmentation with an Atkinson-Shiffrin memory model. ECCV, 2022.

[5] Ke, L. et al. Segment anything in high quality. NeurIPS, 2023.

[6] Liu, Y. et al. Learning quality aware dynamic memory for video object segmentation. ECCV, 2022.

[7] Cheng, H.K., et al. Tracking anything with decoupled video segmentation. ICCV, 2023.

[8] Cheng, H.K., et al. Putting the object back into video object segmentation. CVPR, 2024.

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