

UE Computer Vision 2024

Optical flow based tracking using CNN's

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1.1 Objective

FRAME 1



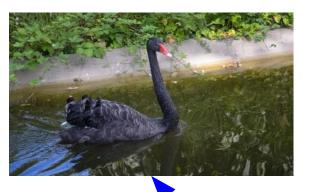
FRAME 2



How to obtain the segmentation of the next frame?

1.2 Operation

1) We compute the inverse sense optical flow (Frame 1 to 2)

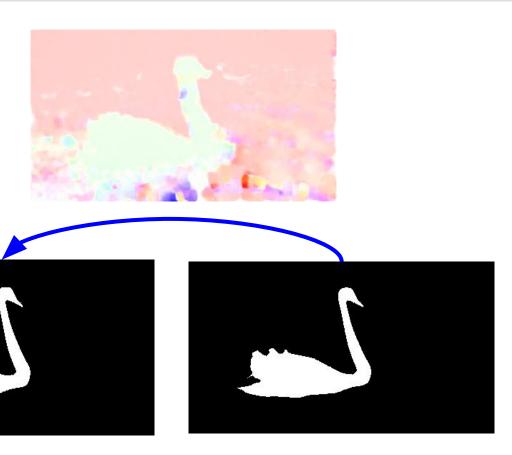






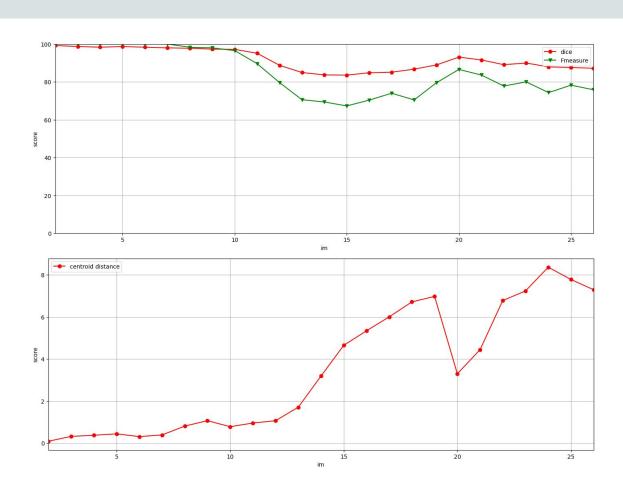
1.2 Operation

2) We propagate the mask by checking if the previous position of each pixel was part of the mask

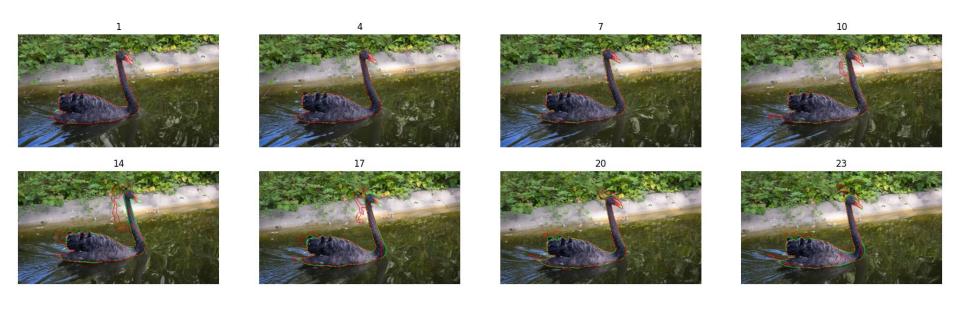


1.3 Performance

This method has a great performance for frames that are close but it degrades with higher time differences.



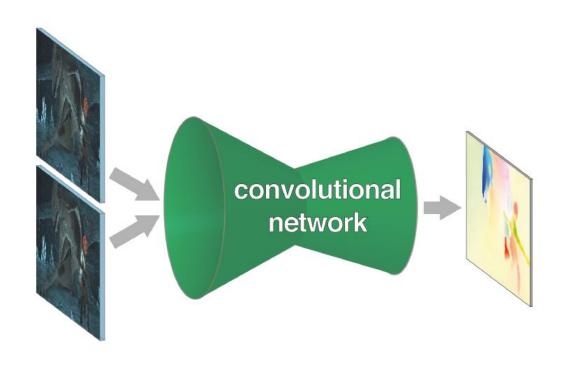
1.3 Performance





2.1 FlowNet recap

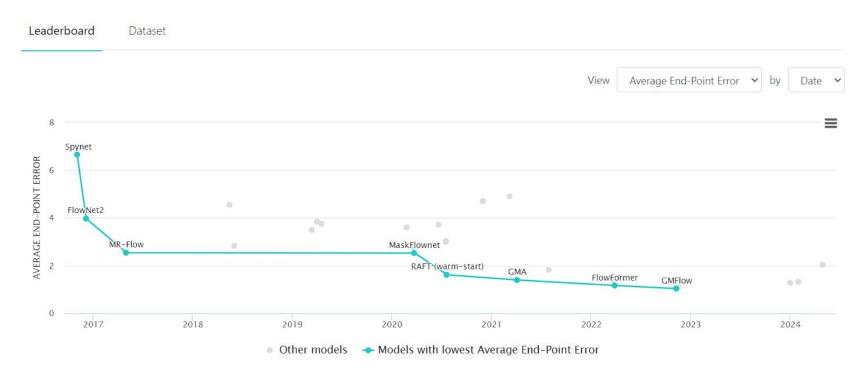
FlowNet consists of a encoder-decoder CNN architecture which is trained to estimate optical flow





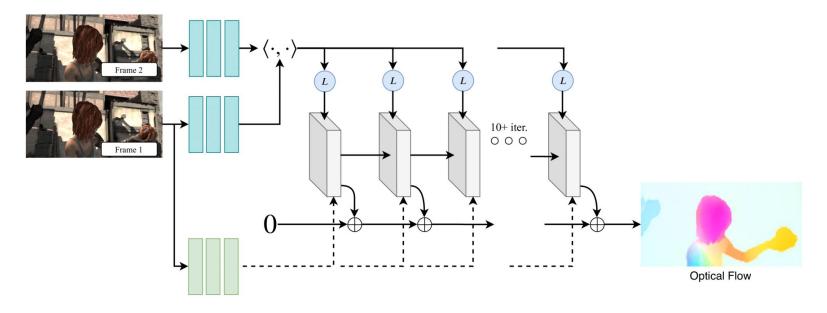
2.2 Current SOTA

Optical Flow Estimation on Sintel-clean



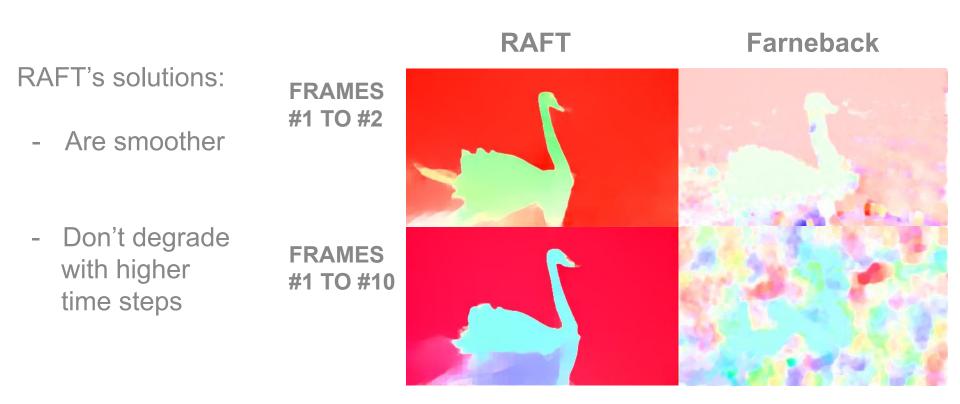
2.3 RAFT

RAFT is a modern CNN based OF model which acts in a recurrent manner.



The output is refined through a certain number of iterations.

2.4 RAFT vs Farneback



3. The Ghost Duck





3.1 RAFT segmentation propagation

Our initial results were not encouraging...





(GIF)

3.1 RAFT segmentation propagation

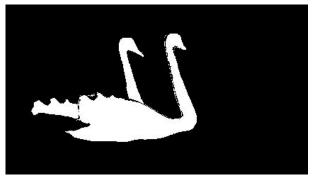
The segmentation propagation method is not suited for RAFT.





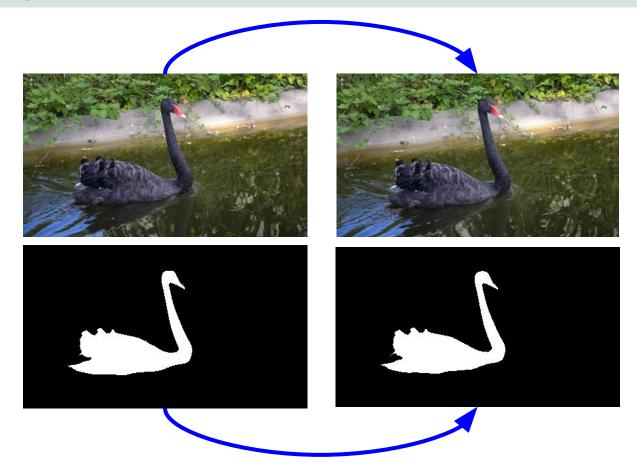
The occluded background is assigned as part of the segmentation





3.1 RAFT segmentation propagation

We solved it by using the forward sense OF and propagating the position of each pixel of the mask.



3.1 RAFT segmentation propagation

Now the ghost is gone, but we need to use some post-processing to correct some issues:

- Cracks
- Small objects



(GIF)

3.1 RAFT segmentation propagation

With morphological operations we remove small objects and fill in the cracks:

- Dilatation
- Erosion
- Remove small object
- Fill

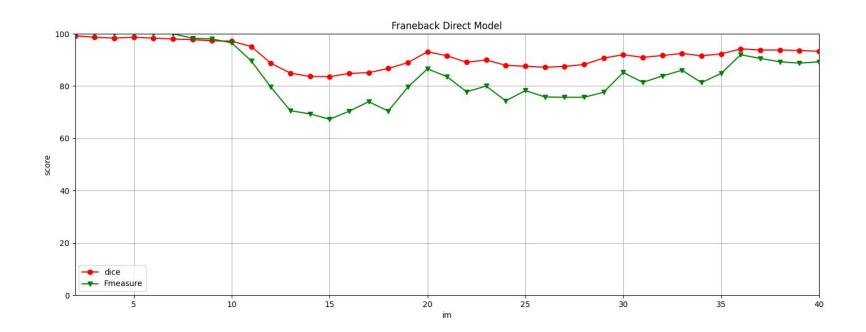


(GIF)

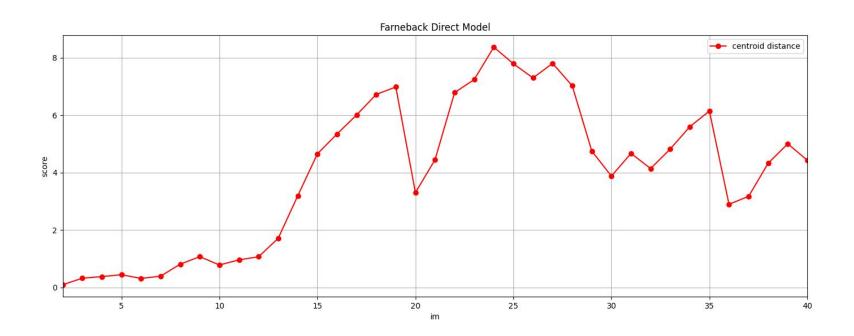
4.Results



Farneback Direct Model

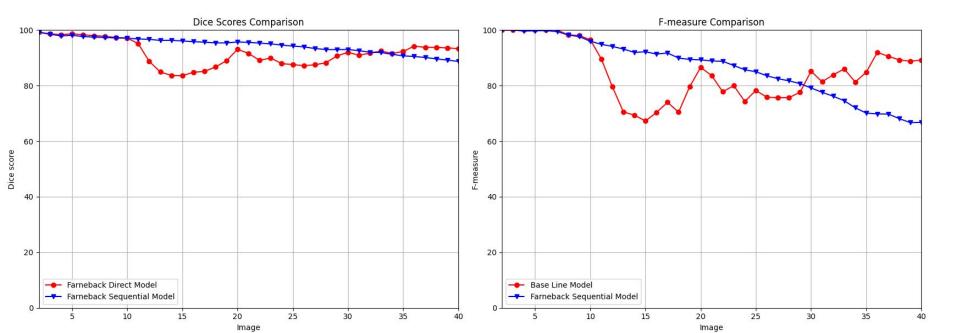


Farneback Direct Model

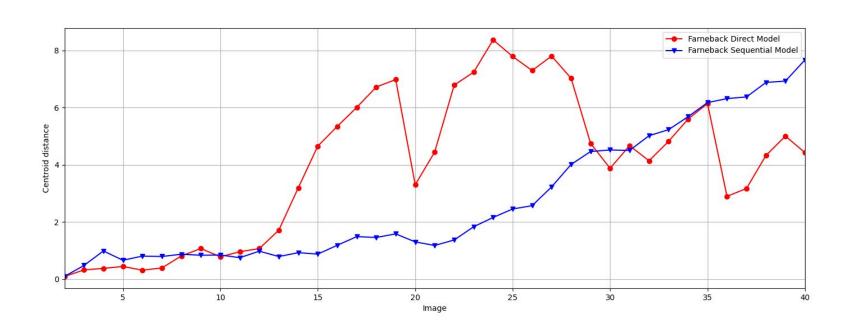


4. Results

Farneback Sequential model

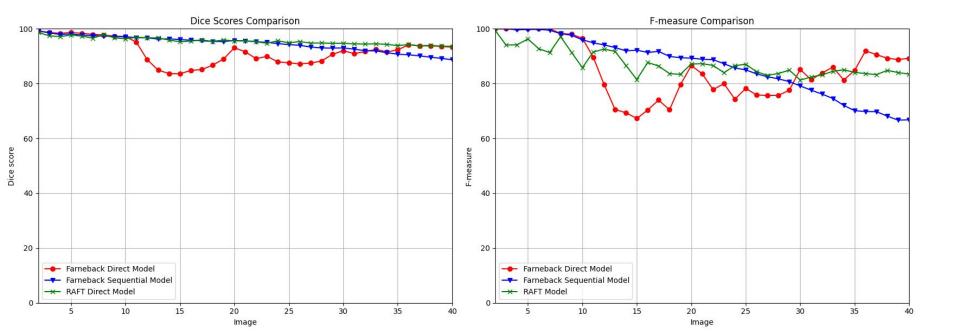


Farneback Sequential model

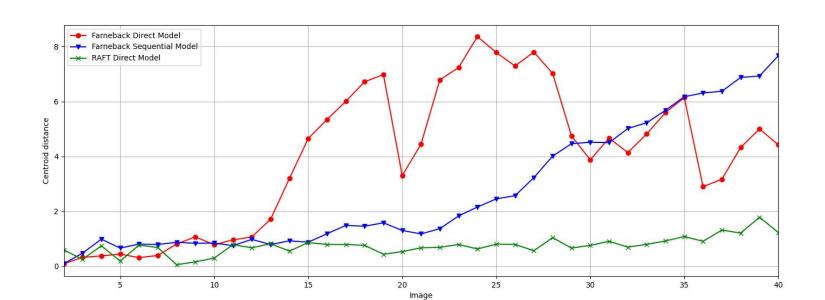


4. Results

RAFT Direct Model

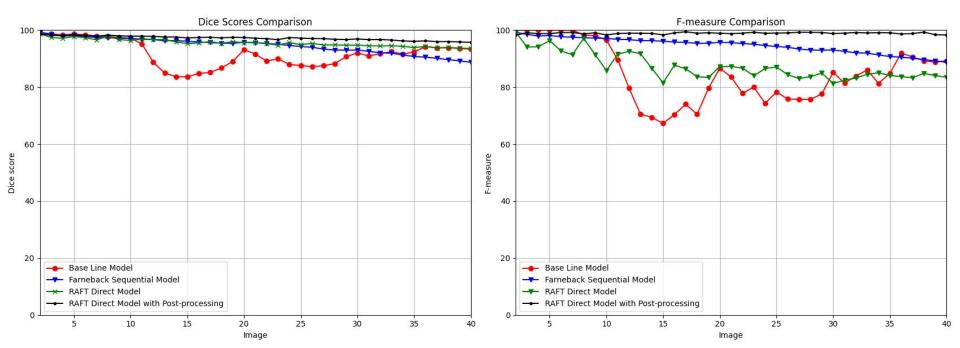


RAFT Direct Model

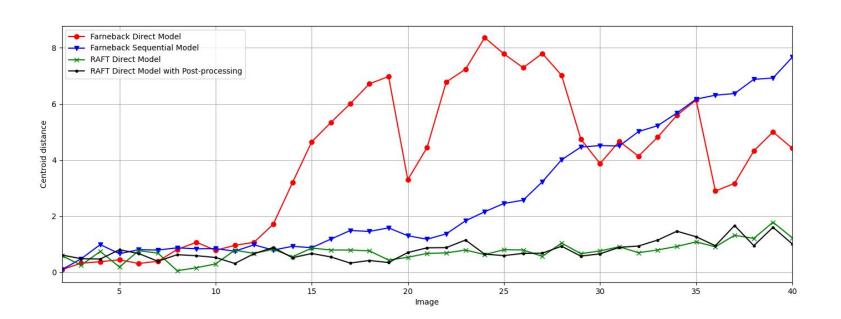


4. Results

RAFT Direct Model with Post-processing



RAFT Direct Model with Post-processing



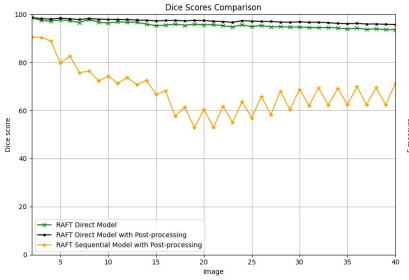
4. Future Improvements

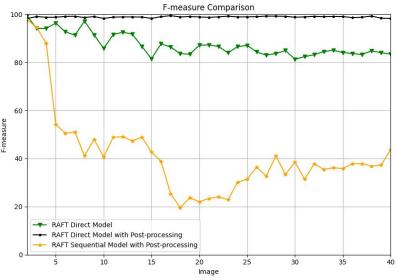


RAFT Sequential Model with Post-processing



Do to a lack of time, we could not make it work correctly



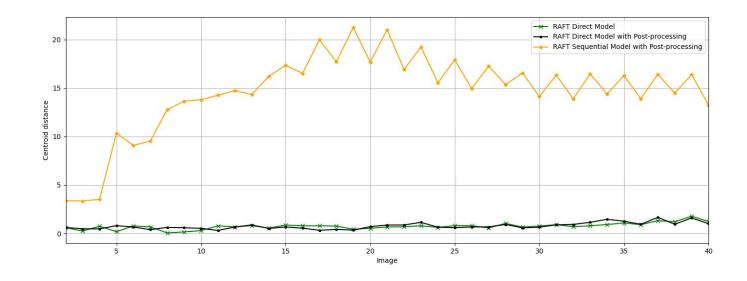




RAFT Sequential Model with Post-processing



Do to a lack of time, we could not make it work correctly

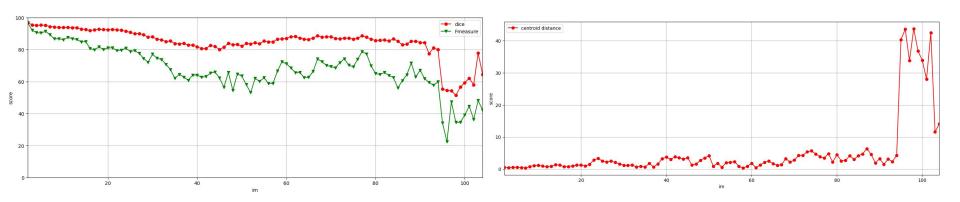


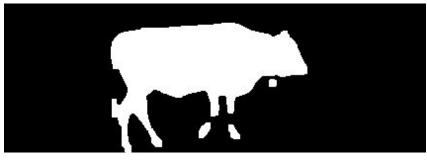


5. Challenge Results



4. Results





(GIF)

Thank you for your attention

