

Transforming a Standard Video into Slow Motion using AI

**A Seminar on Applications of AI by
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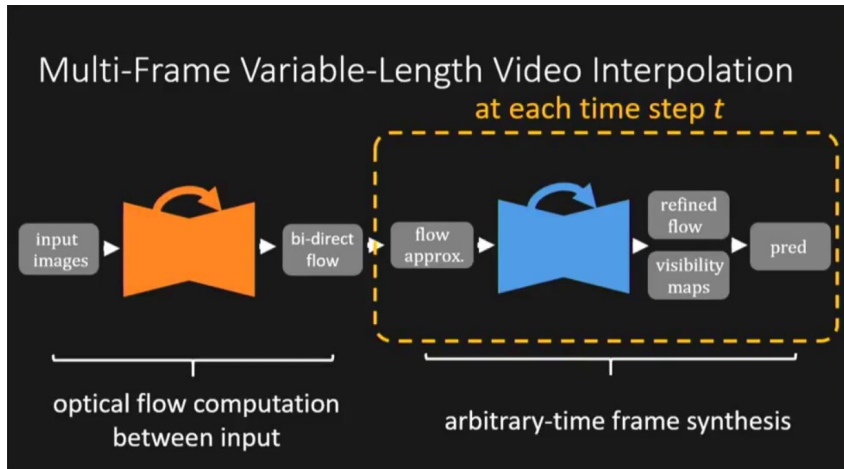
September 19, 2018

Motivation

There are many memorable moments in your life that you might want to record with a camera in slow-motion because they are hard to see clearly with your eyes: the first time a baby walks, a difficult skateboard trick, etc.

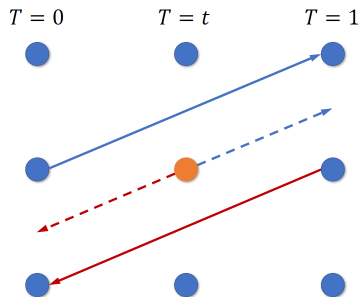
- Professional high-speed cameras are required for higher fps
- Moments we would like to slow down are unpredictable, and as a result, are recorded at standard frame rates.
- Recording everything at high frame rates is **impractical**—it requires large memories and is powerintensive for mobile devices.
- It also has intriguing new applications in self-supervised learning, serving as a supervisory signal to learn optical flow from unlabeled videos.

Multi-frame Variable-Length Video Interpolation



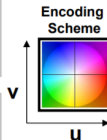
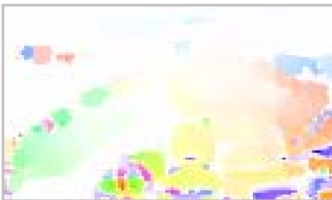
Optical Flow computation

Optical flow or optic flow is the pattern of **apparent motion** of objects, surfaces, and edges in a visual scene caused by the relative motion between an observer and a scene.



The orange pixel borrows optical flow from pixels at the same position in the first and second images.

Optical Flow - Example



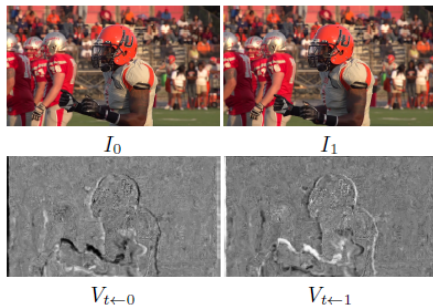
Videos

Color Coded Optical Flows

Refined Flow

- We take this Optical Flow and refine it further
- As approximation works well in smooth regions but poorly around motion boundaries, as the motion near motion boundaries is not locally smooth.
- To **reduce artifacts** around motion boundaries, which may cause poor image synthesis, we propose learning to refine the initial approximation.

Sample of predicted Visibility Maps



- The arms move downwards from $T=0$ to $T=1$. So the area right above the arm at $T=0$ is visible at t but the area right above the arm at $T=1$ is occluded
- The white area around arms in $V_{t \rightarrow 0}$ indicate such pixels in I_0 contribute most to the synthesized I_t while the occluded pixels in I_1 have little contribution.

Synthesizing the Intermediate Frame

- First use both the images (image at time $t=0$ and image at time $t=1$) and the corresponding refined **bi-directional optical flows** to backward warp these images.
- Then use the second neural network which takes these warped images, the visibility maps and fuses them together.



Why this method is the best?

- Produces as many intermediate frames as we want (**time independent**)
- Directly applicable to different scenarios without any modification
- Produces results that are less blurry and more fluidic than other approaches

Current Limitations

- It is not perfect and cannot match natively shot slow motion videos
- Not Opensource and not available to public
- Requires a basic amount of fps and quality of the original video for a legible output



(a)



(b)



(c)



(d)



(e)



(f)

Figure: Visual Comparison with other Techniques

References

- <https://www.youtube.com/watch?v=MjViy6kyiqs>
- <https://news.developer.nvidia.com/transforming-standard-video-into-slow-motion-with-ai/>
- <https://arxiv.org/abs/1712.00080>
- <https://www.youtube.com/watch?v=rkSzRbM4lIM>
- Some Images taken from Google

Thank you!

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