

The background features a soft, yellowish-gold wash with a subtle texture. Scattered across the top and left sides are numerous autumn leaves in shades of orange, red, and yellow, some with green veins. The overall aesthetic is warm and artistic.

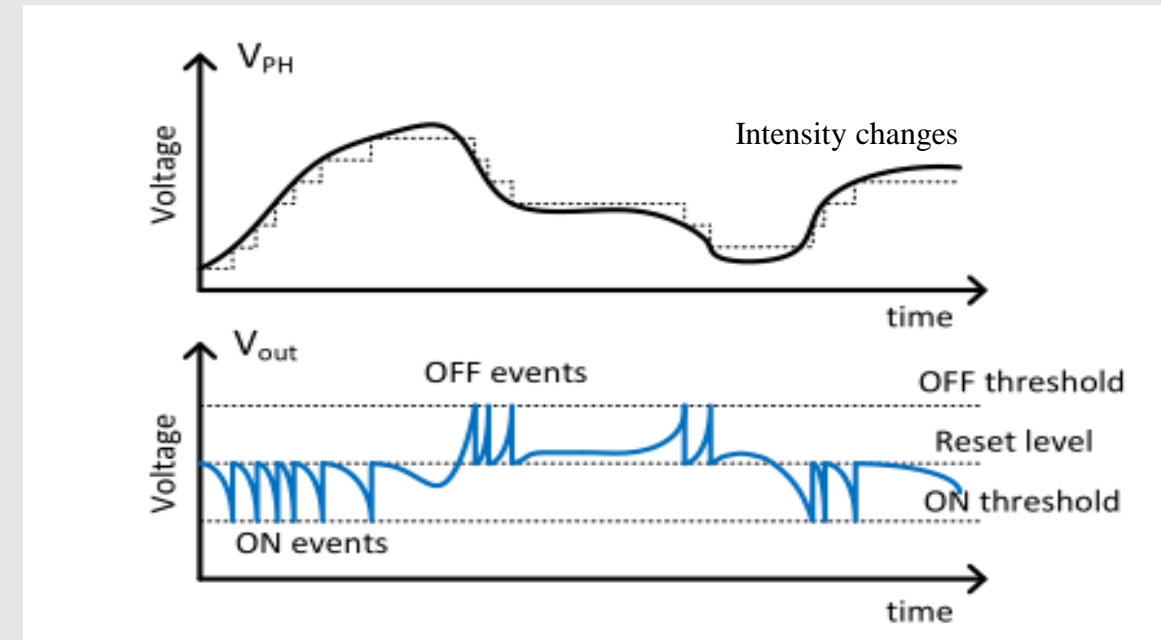
ICASSP 2020

**Intensity-Image Reconstruction For Event Cameras
Using Convolutional Neural Network**

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Introduction

- **What is an event camera?**
- Dynamic Vision Sensor (DVS)
- Pixels measure intensity changes asynchronously and independently
- Event (location X, Y , Polarity, Time)



Generate events at a pixel when intensity changes exceed the threshold

Working principle of event cameras

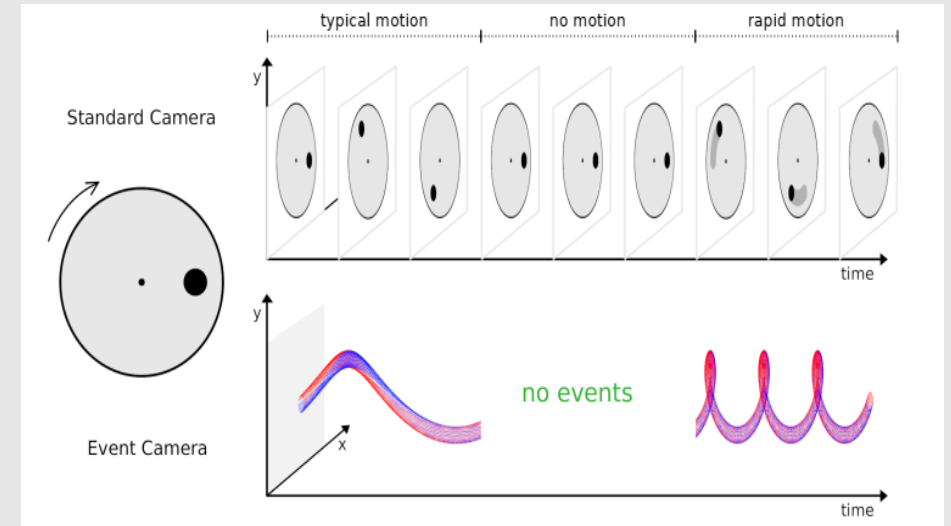
Introduction

Advantages of event cameras

- High dynamic range
- No motion blur
- low redundancy
- Low latency (μs)
- High temporal resolution (μs)

Purpose of research:

- Advantages of using event cameras.
- Compatible with existing frame-based algorithms
- Screen display and human observation



Event cameras VS standard cameras



HDR

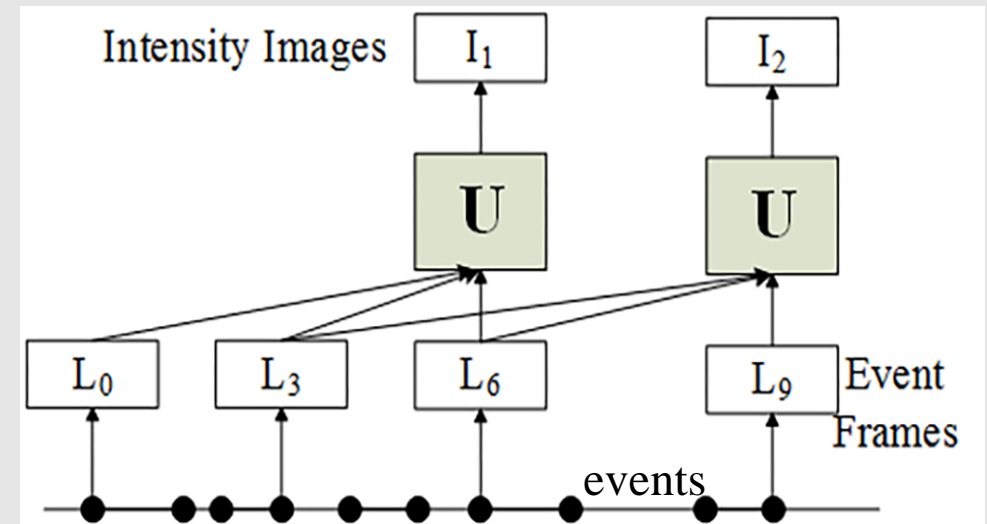
Method

Two steps

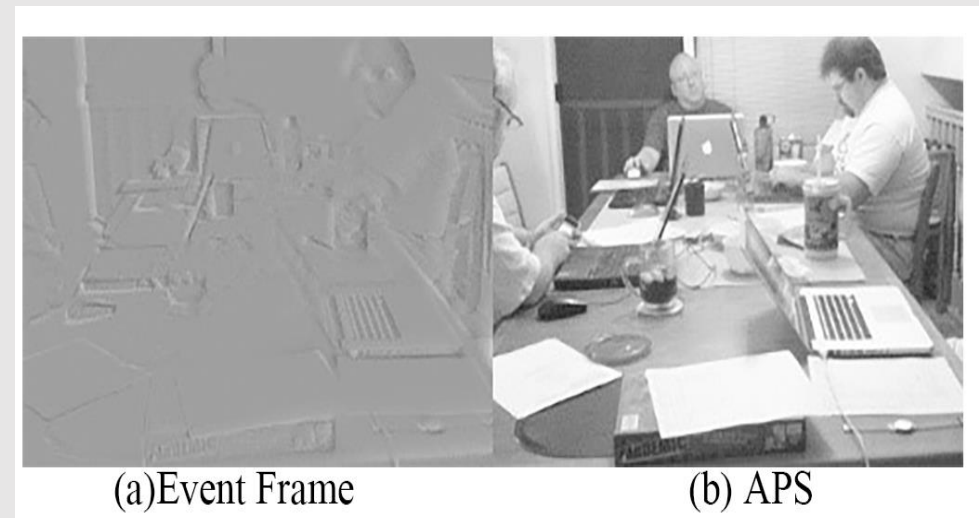
- Convert the events into event frames
- U-net (U) network transforms event frames into intensity images (I).

Some details

- The U-net is trained on simulated datasets.
- The simulator ESIM generate event streams and time-stamped frame images(APS).
- Perceptual loss



Overview of our approach



A pair of event frame (one channel of the input image) and APS image (the ground truth) in the training dataset.

Method

An attenuation method to generate event frames.

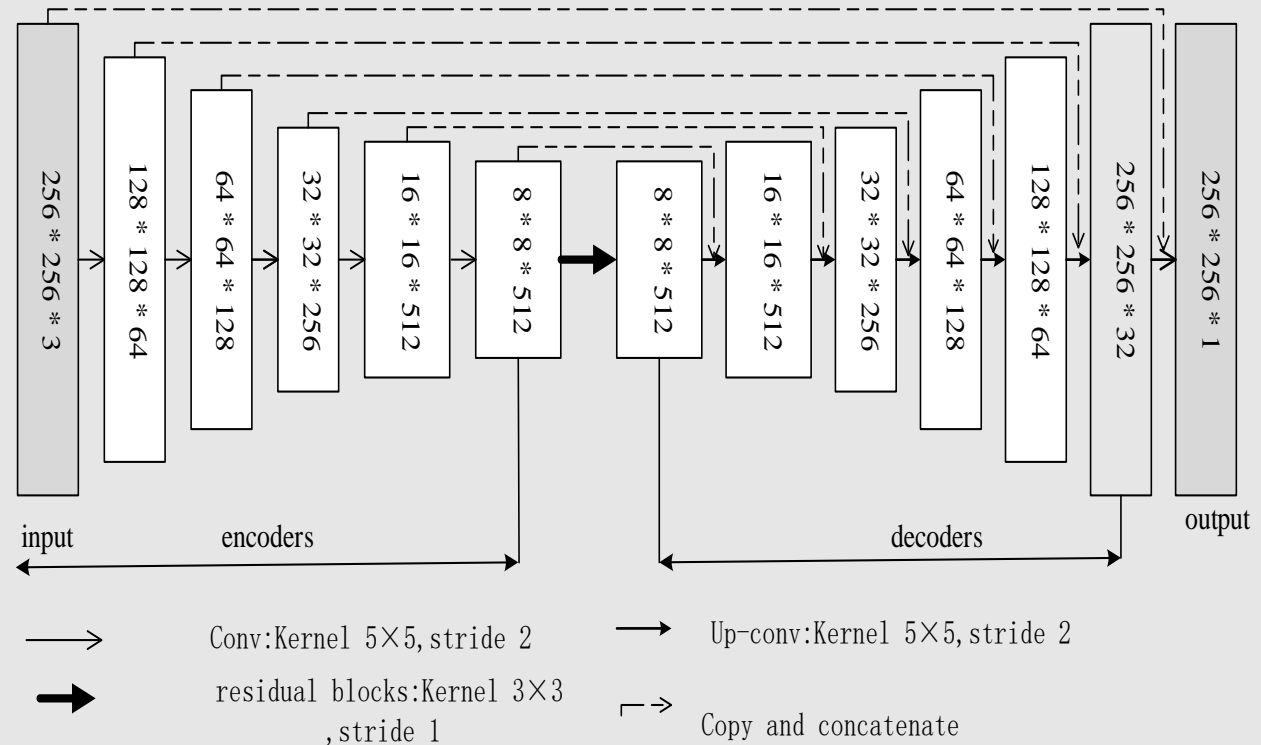
- If the same pixel (x, y) generates several events $e_0, e_1, e_2 \dots e_k \dots e_n$.
- The threshold of e_k is expressed as C_k .
- $L_k(x, y)$ is the intensity value of e_k
- β is an attenuation coefficient less than 1.
- $L_k(x, y)$ is affected by all events that are generated by pixel (x, y)
- The magnitude of influence is reflected in β^n .
- Historical information is attenuated.

$$\begin{cases} L_0(x, y) = C_0 \\ L_1(x, y) = L_0(x, y) \times \beta + C_1 \\ L_k(x, y) = L_{k-1}(x, y) \times \beta + C_k \\ L_n(x, y) = L_{n-1}(x, y) \times \beta + C_n \\ = C_0 \times \beta^n + C_1 \times \beta^{n-1} + \dots C_k \times \beta^{n-k} + \dots + C_n \times \beta^0 \end{cases}$$

Method

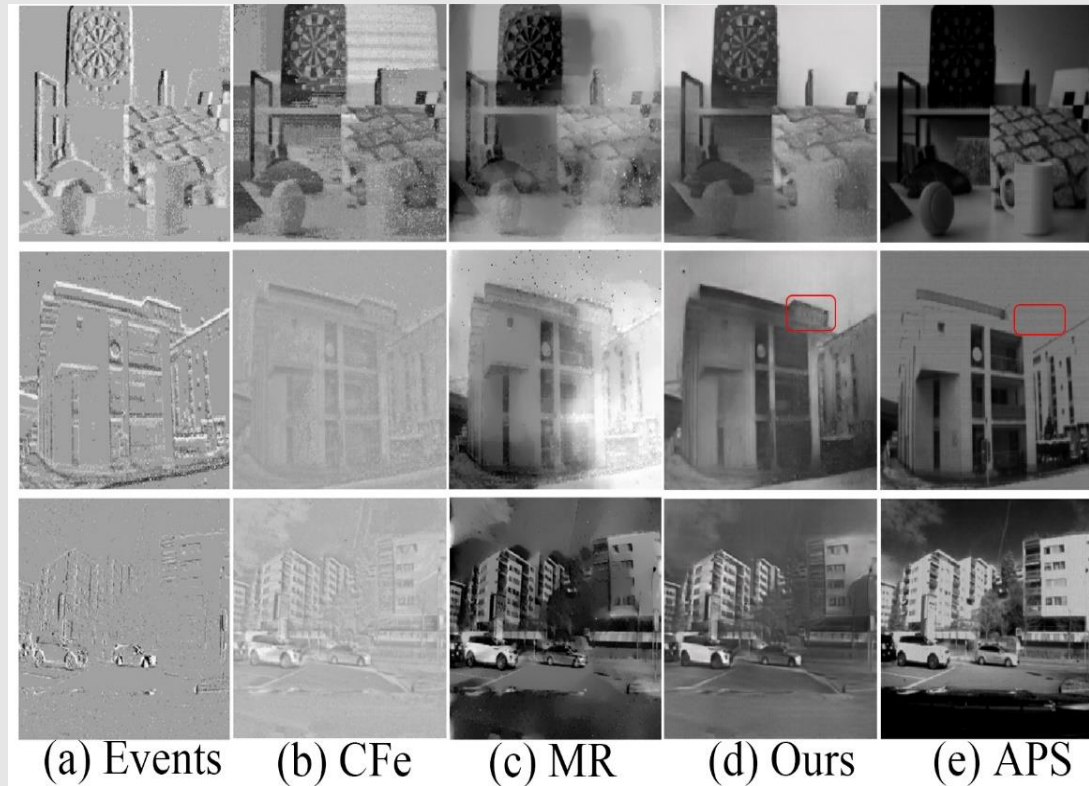
U-net network

- The network has 5 encoders, 1 residual block, 5 decoders and a final convolutional layer
- Skip connections



U-net

Experiments and Evaluation

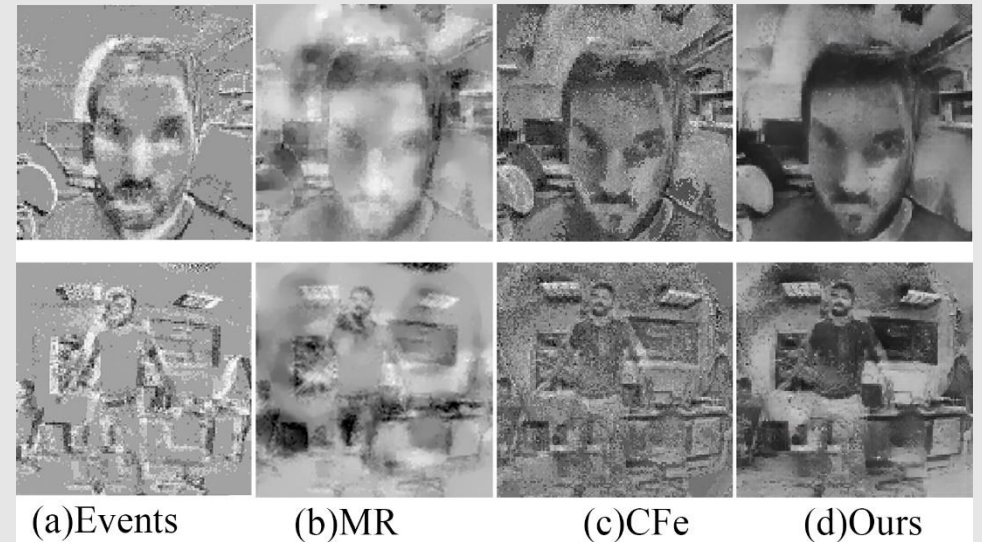


Generated images of different model. The column (a) shows events, and the column (e) shows APS images of DAVIS for reference, and other columns show the generated images.

Experiments and Evaluation

Table 1 *quantitative comparison using SSIM and FSIM*

Model	SSIM	FSIM
CFe	0.55	0.68
MR	0.64	0.71
Ours	0.78	0.89



Result in noise dataset.

The background features a soft, warm watercolor wash in shades of yellow and orange. Scattered throughout are numerous autumn leaves in various colors, including red, orange, yellow, and some green, creating a seasonal and artistic atmosphere.

Thanks for listening !

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