



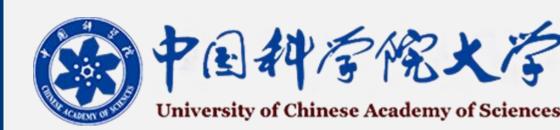
# Optical Flow for Spike Camera with Hierarchical Spatial-**Temporal Spike Fusion**

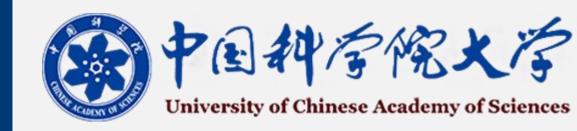




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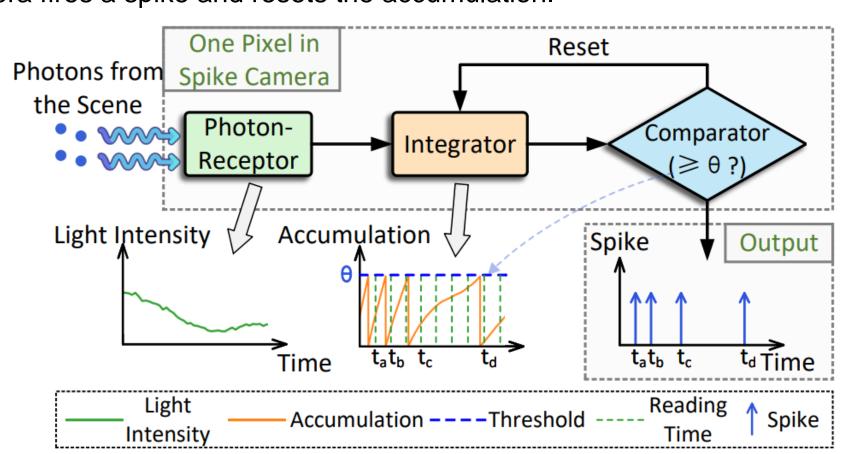




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## 1. Introduction

1.1 Spike Camera. Spike cameras are composed of an array of pixels working asynchronously. Each pixel of a spike camera is composed of three main components: photon-receptor, integrator, and comparator. The integrator accumulates the photoelectrons from the photon-receptor and transfers them to the voltage. The comparator compares the accumulation with the threshold continuously. Once the voltage of the integrator exceeds a certain threshold, the camera fires a spike and resets the accumulation.

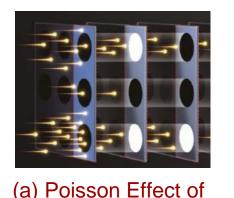


Key components of a pixel in spike camera

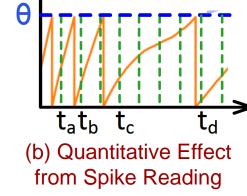
$$A(\mathbf{x}, t) = \int_0^t \alpha \cdot I(\mathbf{x}, \tau) d\tau \mod \theta$$

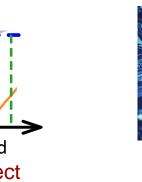
### 1.2 Challenges of Spike-Based Optical Flow.

Noises in the imaging of spike cameras.



Photons' Arrival







(c) Thermal Noises in the Circuits

## Fluctuations and Randomness in Spikes

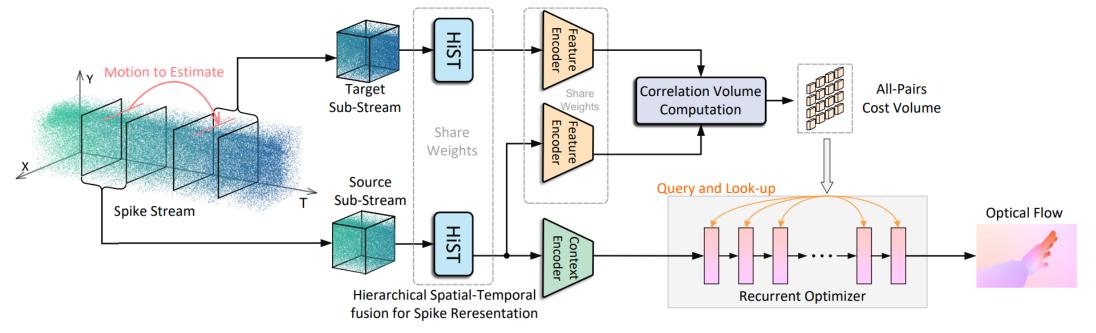
Ambiguities in correlation -> Inaccurate feature matching

## 2. Contributions

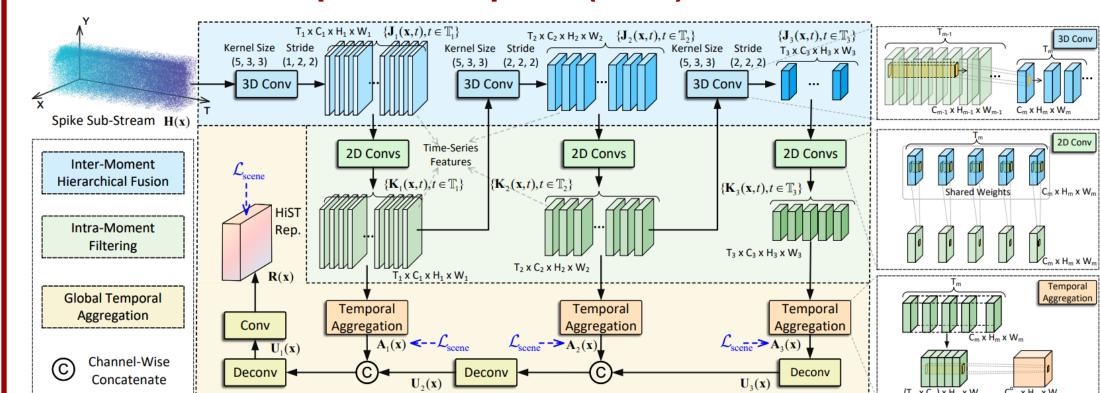
- A HiST-SFlow is proposed for spike-based optical flow. In HiST-SFlow, the spikes are represented by the HiST module and extracted to features for correlation. The optical flow is estimated by a recurrent optimizer.
- An inter-moment hierarchical fusion (InterF) module and an intra-moment filtering (IntraF) module are proposed to suppress the randomness in the spikes. A scene loss is proposed to constrain high-fidelity representation to contain the brightness information of the scene.

## 3. Approaches

#### 3.1 Overall Architecture of HiST-Sflow.



## 3.2 Hierarchical Spatial-Temporal (HiST) Fusion.



#### (a) Inter-Moment Hierarchical Fusion (InterF).

■ Fuse features at different moments while retaining the time information in features.

$$\mathbf{J}_{m}(\mathbf{x},t) = \mathscr{J}_{m} \left[ \left\{ \mathbf{K}_{m-1}(\mathbf{x},\tau) \mid \tau \in \mathbb{T}_{m-1} \right\} \right]$$
$$\mathbb{T}_{m-1} = \left\{ T_{c} - T_{m-1}^{\text{half}}, \dots, T_{c}, \dots T_{c} + T_{m-1}^{\text{half}} \right\}$$

#### (b) Intra-Moment Filtering (IntraF).

- Reduce the influence of spikes' fluctuations for each moment through the feature at the current moment.
- The InterF and IntraF are implemented alternatively in each level of the pyramid.

$$\mathbf{K}_m(\mathbf{x},t) = \mathscr{K}_m[\mathbf{J}_m(\mathbf{x},t)], t \in \mathbb{T}_m$$

#### (c) Global Temporal Aggregation (GTA).

■ Fuse features of all the moments at each level of the pyramid to represent the central moment of input spike sub-stream.

$$\mathbf{A}_m(\mathbf{x}) = \mathscr{A}_m \left[ \operatorname{Cat} \left\{ \mathbf{K}_m(\mathbf{x}, \tau) \mid \tau \in \mathbb{T}_m \right\} \right]$$

#### (d) Scene Loss.

- Ensure the spike representation contain the scene's brightness information
- The  $\{\mathscr{P}_m\}_{m=0}^3$  are 3-layer convolution layers, which are used only during training and not in inference.

$$\mathcal{L}_{\text{scene}} = \|\mathbf{I}_{\text{scene}}(\mathbf{x}, T_{\text{c}}) - \mathscr{P}_0(\mathbf{R}_{T_{\text{c}}}(\mathbf{x}))\|_1 + \sum_{m=1} \lambda_m \|\sigma_m(\mathbf{I}_{\text{scene}}(\mathbf{x}, T_{\text{c}})) - \mathscr{P}_m(\mathbf{A}_m(\mathbf{x}))\|_1$$

#### 3.3 Loss Function.

$$\mathcal{L}_{\text{flow}} = \sum_{i=1}^{N} \gamma^{N-i} \|\mathbf{w}_i(\mathbf{x}) - \mathbf{w}_{\text{gt}}(\mathbf{x})\|_1$$

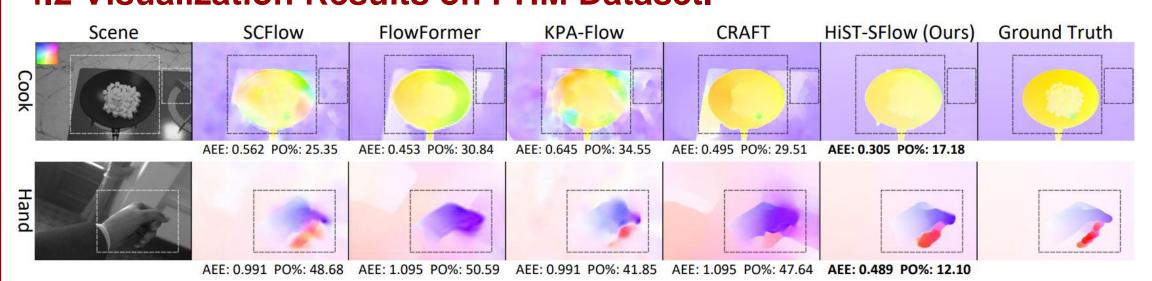
$$\mathcal{L} = \mathcal{L}_{\text{flow}} + \lambda (\mathcal{L}_{\text{scene}}^{\text{src}} + \mathcal{L}_{\text{scene}}^{\text{tgt}})$$

## 4. Experimental Results

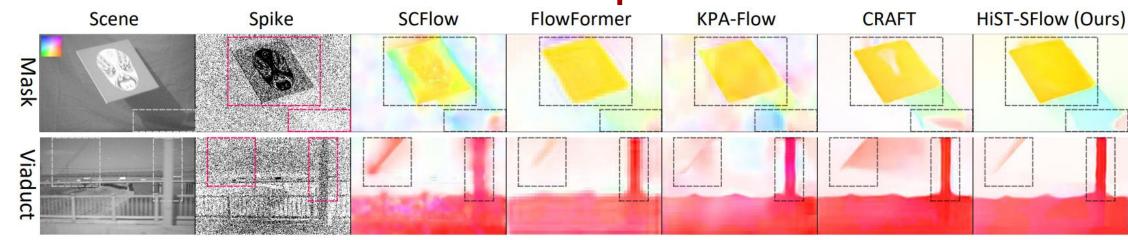
### 4.1 Quantitative Results on PHM Dataset (AEPE / PO%).

Architecture	Ban	Cook	Dice	Doll	Fan	Hand	Jump	Poker	Тор	Average
SCFlow	0.51 / 20.3	1.34 / 38.6	1.10 / 30.7	0.22 / 5.6	<b>0.24</b> / 10.7	1.30 / 57.3	0.11 / 3.0	0.80 / 41.1	2.14 / 17.7	0.863 / 25.00
RAFT	0.46 / 12.5	1.32 / 43.7	0.95 / 29.3	0.24 / 6.7	0.28 / 12.7	1.11 / 45.1	0.11 / 3.0	0.67 / 37.1	2.19 / 19.7	0.813 / 23.30
GMA	0.61 / 21.7	1.84 / 74.7	1.13 / 34.2	0.39 / 9.4	0.36 / 12.1	2.13 / 80.6	0.17 / 2.8	0.88 / 43.5	2.29 / 23.6	1.087 / 33.63
⊆ Flow1D	0.79 / 51.4	1.28 / 50.8	1.15 / 47.9	0.27 / 6.3	0.28 / 11.0	1.86 / 83.1	0.13 / 3.4	0.85 / 50.1	2.19 / 17.7	0.979 / 35.76
KPA-Flow	0.47 / 14.9	1.41 / 45.9	0.87 / 29.9	0.27 / 7.1	0.29 / 12.7	1.19 / 47.7	0.12 / 3.0	0.65 / 36.6	2.19 / 19.4	0.827 / 24.12
GMFlow	0.76 / 42.4	1.29 / 61.0	1.54 / 81.7	0.31 / 8.4	0.43 / 14.1	1.83 / 65.0	0.30 / 3.7	0.95 / 54.2	2.29 / 23.3	1.077 / 39.33
GMFlowNet	0.45 / 12.1	1.22 / 43.8	1.02 / 32.9	0.35 / 7.8	0.25 / 10.7	1.53 / 65.3	0.12 / 3.2	0.65 / 31.5	2.18 / 17.5	0.863 / 24.98
CRAFT	0.61 / 15.0	1.28 / 43.5	0.93 / 27.6	0.19 / 5.0	0.25 / 10.2	1.67 / 73.3	0.10 / 2.6	0.56 / 23.1	2.15 / 15.1	0.860 / 23.94
FlowFormer	0.52 / 13.5	1.48 / 58.7	0.98 / 31.0	0.25 / 6.7	0.29 / 11.5	1.82 / 84.5	0.14 / 3.6	0.94 / 54.9	2.22 / 19.5	0.959 / 31.54
HiST-SFlow	0.28 / 7.8	0.80 / 27.4	0.85 / 23.3	0.20 / 5.6	0.27 / 12.8	0.64 / 21.7	0.08 / 2.5	<b>0.53</b> / 23.9	2.11 / 14.8	0.640 / 15.54
SCFlow	0.94 / 27.1	3.00 / 50.6	1.72 / 33.2	0.41 / 8.1	0.46 / 13.6	3.71 / 71.3	0.19 / 5.9	1.57 / 53.7	4.25 / 18.9	1.804 / 31.37
RAFT	0.78 / 18.6	2.75 / 54.4	1.57 / 30.1	0.43 / 9.3	0.50 / 14.6	2.81 / 59.9	0.21 / 5.8	1.31 / 46.7	4.30 / 21.2	1.628 / 28.94
GMA	1.01 / 22.1	4.95 / 96.4	1.52 / 35.9	1.00 / 59.6	1.19 / 98.4	6.66 / 99.5	0.81 / 84.4	1.39 / 45.2	4.64 / 64.9	2.575 / 67.38
⊖ Flow1D	1.19 / 51.6	4.52 / 96.3	1.58 / 50.7	0.78 / 53.3	1.01 / 82.1	6.65 / 99.2	0.72 / 73.1	1.39 / 52.3	4.75 / 79.7	2.510 / 70.90
KPA-Flow	0.80 / 20.9	2.93 / 55.6	1.48 / 31.4	0.45 / 9.6	0.52 / 14.5	2.86 / 62.5	0.22 / 5.6	1.31 / 48.4	4.28 / 19.7	1.649 / 29.81
♣ GMFlow	1.49 / 80.3	2.64 / 80.1	2.72 / 91.8	0.54 / 15.3	0.77 / 22.0	3.79 / 81.5	0.55 / 27.8	1.78 / 75.3	4.45 / 32.5	2.080 / 56.28
GMFlowNet	0.92 / 31.4	2.61 / 70.4	2.17 / 42.7	0.61 / 27.5	0.56 / 13.9	3.30 / 93.2	0.21 / 4.5	1.33 / 53.4	4.33 / 25.3	1.782 / 40.25
CRAFT	1.16 / 85.5	2.68 / 61.0	1.99 / 46.8	0.39 / 7.8	0.48 / 12.5	3.53 / 87.1	0.20 / <b>3.6</b>	<b>1.23</b> / 38.9	4.31 / 22.0	1.775 / 40.57
FlowFormer	0.91 / 13.8	4.41 / 96.3	<b>1.40</b> / 32.6	0.80 / 54.8	1.03 / 90.0	6.54 / 99.3	0.74 / 75.8	1.47 / 57.4	4.59 / 61.9	2.432 / 64.67
HiST-SFlow	0.55 / 8.8	2.04 / 33.6	1.64 / <b>26.3</b>	0.38 / 7.2	0.51 / 13.9	2.00 / 34.7	<b>0.17</b> / 5.0	1.28 / <b>33.1</b>	4.18 / 15.1	1.417 / 19.73

#### 4.2 Visualization Results on PHM Dataset.



#### 4.3 Visualization Results on Real-Captured Data.



#### 4.4 Ablation Studies.

#### **Ablations on Proposed Modules**

Index		Settings		$\Delta t$ =	= 10	$\Delta t = 20$		
1110011	InterF	IntraF	$\mathcal{L}_{ ext{scene}}$	AEPE	PO%	AEPE	PO%	
(A)	×	X	X	0.986	33.17	2.095	56.56	
(B)	✓	X	X	0.694	18.18	1.449	21.99	
(C)	✓	$\checkmark$	X	0.676	17.34	1.433	22.79	
(D)	✓	X	$\checkmark$	0.675	16.63	1.448	21.40	
(E)	✓	✓	✓	0.640	15.54	1.417	19.73	

#### **Ablations on Different Representations**

Representation	$\Delta t$ =	= 10	$\Delta t = 20$		
	AEPE	PO%	AEPE	PO%	
Window-Based	0.868	25.72	1.757	34.19	
Interval-Based	0.880	29.77	1.824	37.91	
Multi-Window	0.799	21.10	1.703	34.58	
Flow-Guided Window	0.696	16.99	1.533	23.36	
HiST (Ours)	0.640	15.54	1.417	19.73	

## 4.5 Using HiST for Other Baselines.

Architecture	with HiST	$\Delta t$ =	= 10	$\Delta t = 20$		
	***************************************	AEPE	PO%	AEPE	PO%	
GMA	No	1.087	33.63	2.575	67.38	
GMA	Yes	0.666	16.91	1.391	21.20	
KPA-Flow	No	0.827	24.12	1.649	29.81	
	Yes	0.659	16.99	1.363	22.27	
GMFlowNet	No	0.863	24.98	1.782	40.25	
GMFIOWNEL	Yes	0.730	21.22	1.452	24.93	

