

# RWF-2000: An Open Large Scale Video Database for Violence Detection

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# OUTLINE

1. Introduction

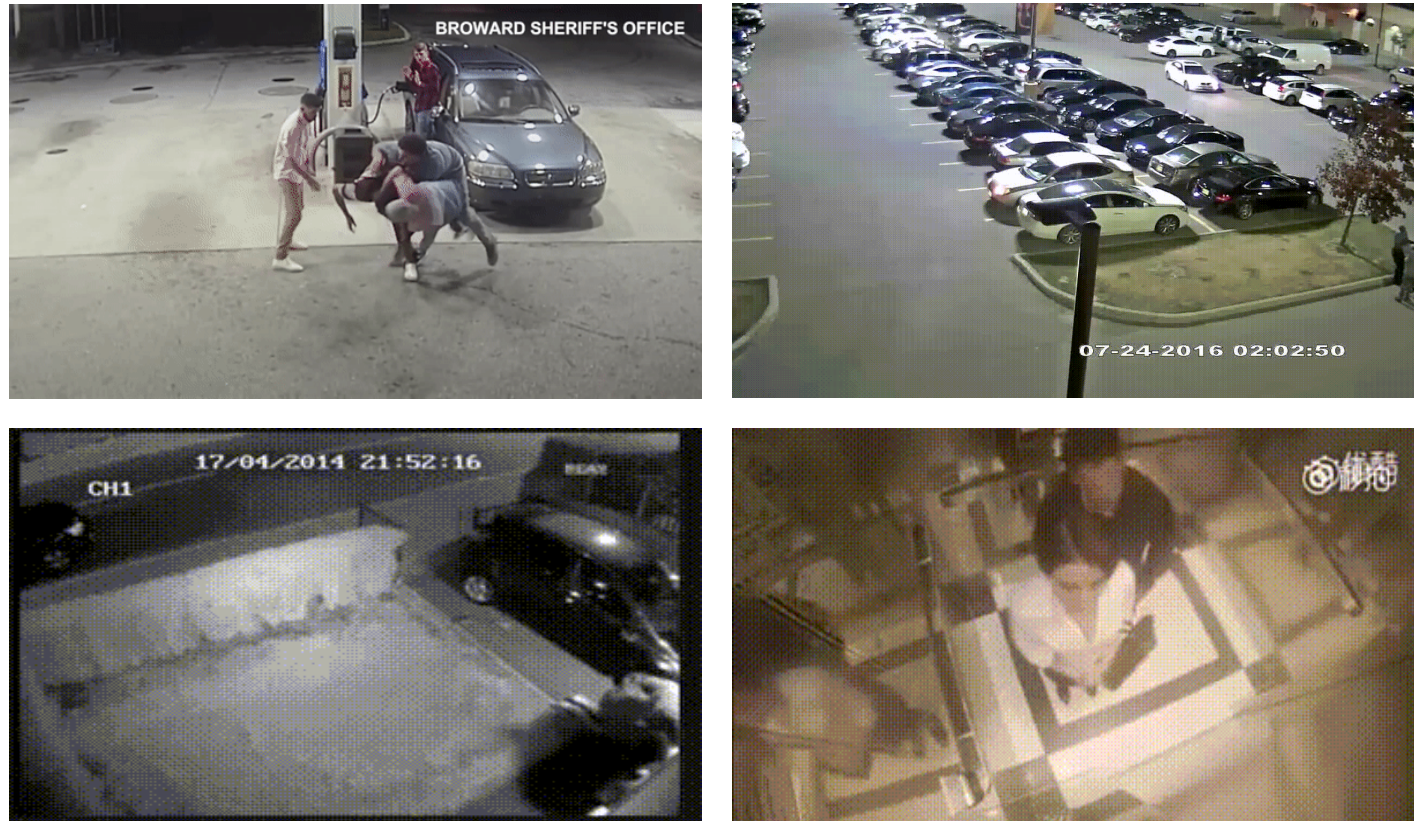
2. Datasets

3. Methods

4. Experiments

5. Demonstration

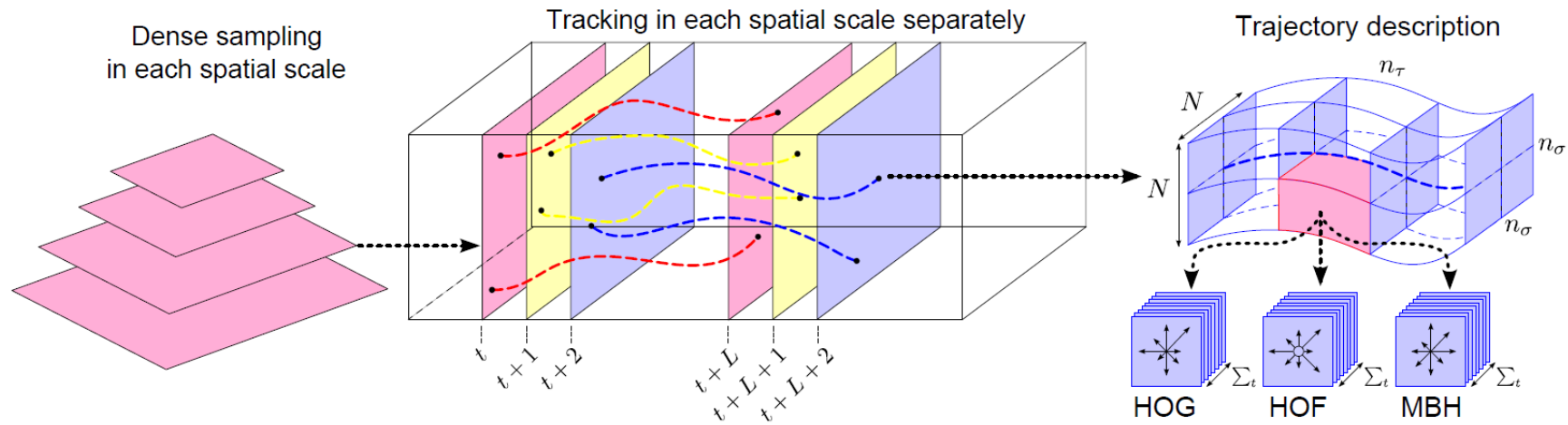
# 1.1 Motivation



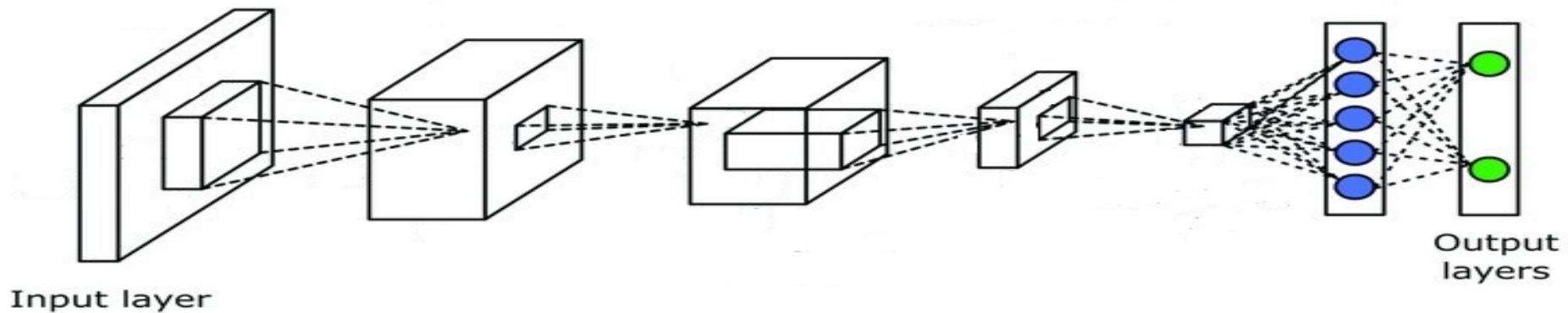
**Fig 1. Violent Activities in cities**

- Surveillance cameras just provide cues and evidences after crimes have been conducted.
- It is both time and labor consuming to manually monitor the large amount of video data.
- Automatically recognizing violence becomes important.

# 1.2 Related Work



**Fig 2. Traditional Method**



**Fig 3. Deep Learning based Method**



# 2.1.1 Previous Datasets



**Crowd Violence**

246 videos captured in crowded places



**Movies Fight**

200 videos extracted from action movies



**Hockey Fight**

1k videos extracted from hockey games

**Fig 4. Previous Datasets**

## 2.1.2 Previous Datasets

Table I

COMPARISONS BETWEEN THE RWF-2000 AND THE PREVIOUS DATASETS. THE 'NATURAL' REPRESENTS THAT VIDEOS ARE FROM REALISTIC SCENES, BUT RECORDED BY HYBRID TYPES OF DEVICES (E.G., MOBILE CAMERAS, CAR-MOUNTED CAMERAS).

Authors	Dataset	Data Scale	Length/Clip (sec)	Resolution	Annotation	Scenario
Blunsden et al. [15]	BEHAVE	4 Videos (171 Clips)	0.24-61.92	640×480	Frame-Level	Acted Fights
Rota et al. [16]	RE-DID	30 Videos	20-240	1280×720	Frame-Level	Natural
Demarty et al. [17]	VSD	18 Movies (1,317 Clips)	55.3-829.4	Variable	Frame-Level	Movie
Perez et al. [18]	CCTV-Fights	1,000 clips	5-720	Variable	Frame-Level	Natural
Nievas et al. [4]	Hockey Fight	1,000 Clips	1.6-1.96	360×288	Video-Level	Hockey Games
Nievas et al. [5]	Movies Fight	200 Clips	1.6-2	720×480	Video-Level	Movie
Hassner et al. [6]	Crowd Violence	246 Clips	1.04-6.52	Variable	Video-Level	Natural
Yun et al. [19]	SBU Kinect Interaction	264 Clips	0.67-3	640×480	Video-Level	Acted Fights
Sultani et al. [20]	UCF-Crime	1,900 Clips	60-600	Variable	Video-Level	Surveillance
Ours	RWF-2000	2,000 Clips	5	Variable	Video-Level	Surveillance



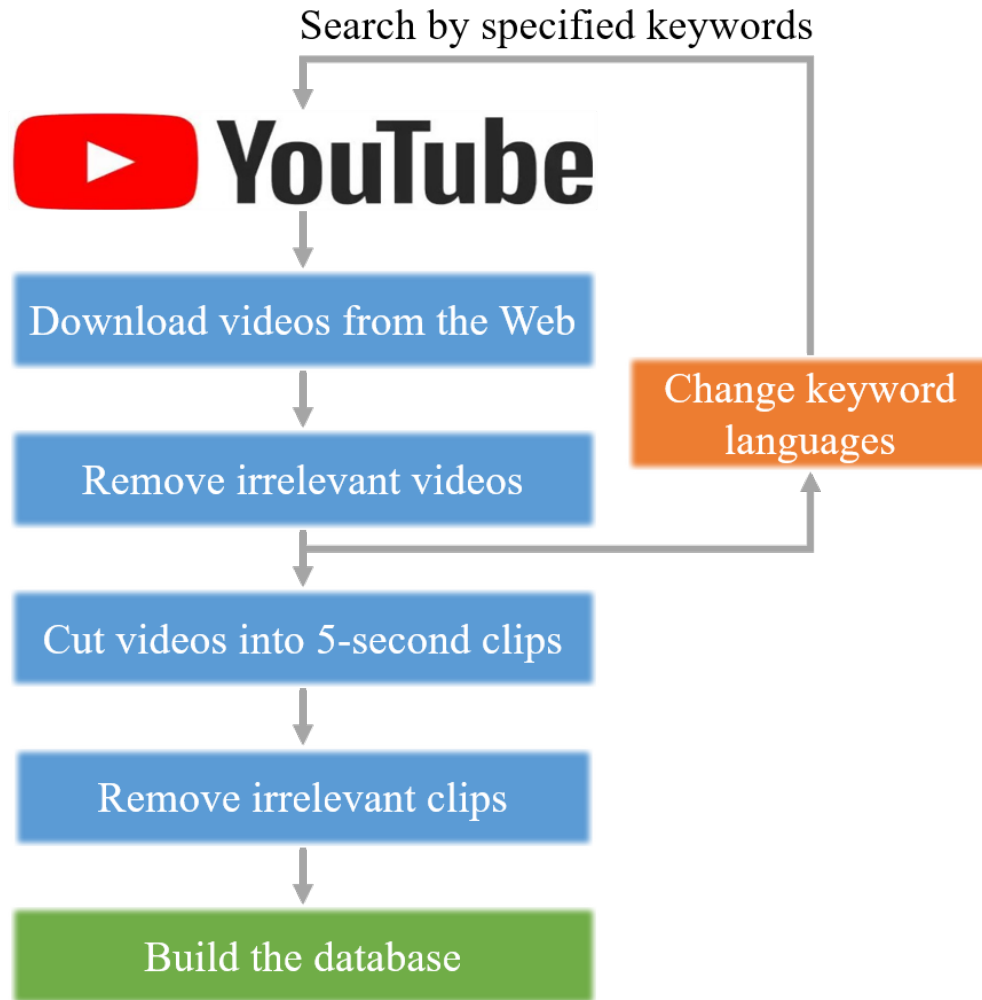
# 2.2.1 Proposed Dataset



**Fig 5. RWF-2000 Dataset**

2000 real-world videos captured by surveillance cameras, with large diversity

## 2.2.2 Proposed Dataset



**Fig 6. Pipeline of Data Collection**

### **Collections of the RWF-2000**

- Search and download videos from the YouTube website
- Remove irrelevant contents and cut videos into clips
- Repeat the above procedures by changing specified keywords
- Annotate collected clips manually to build the database



## 2.2.3 Proposed Dataset

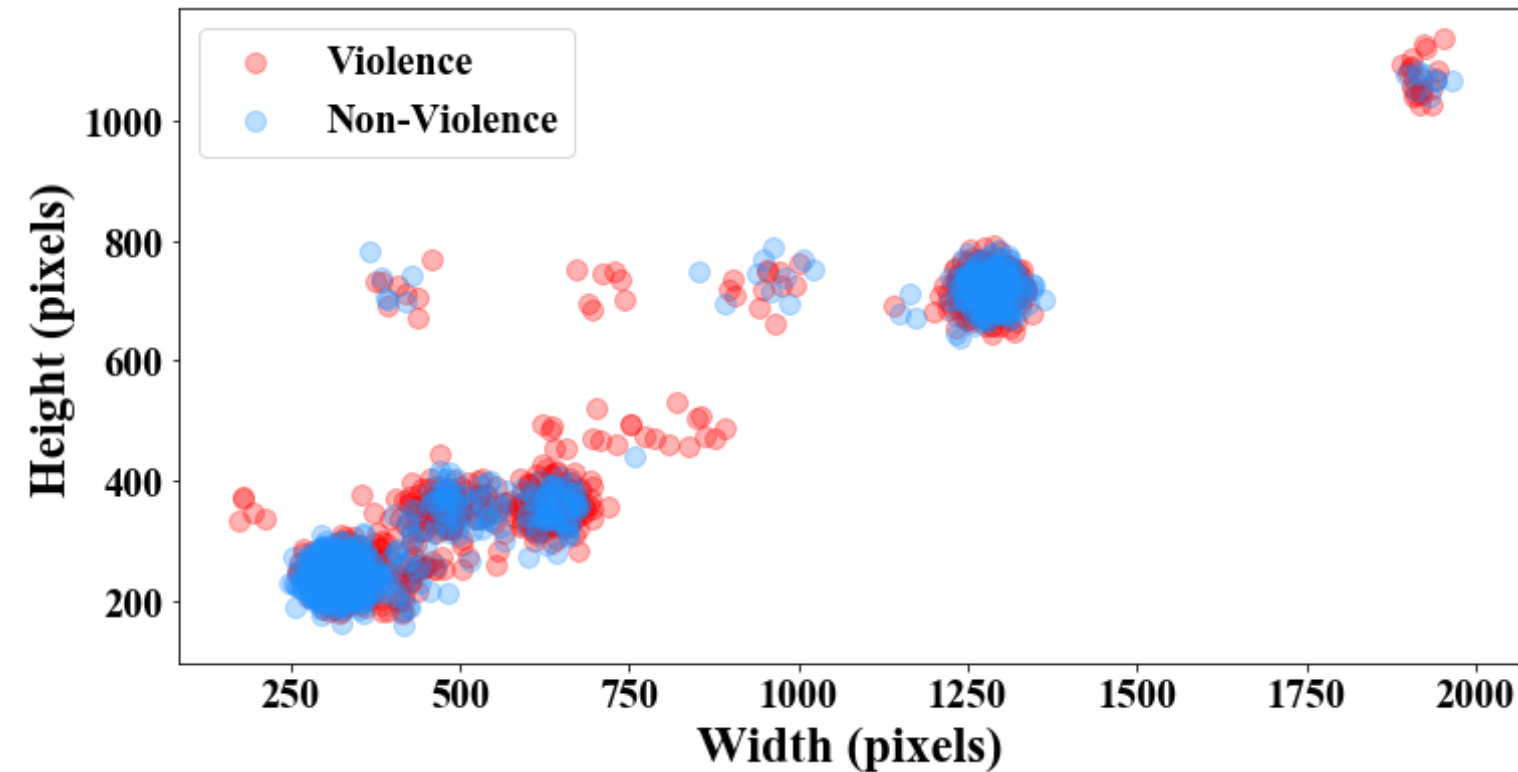


Fig 7. Resolution Distribution of the RWF-2000

### Properties of the RWF-2000

- Large diversity
- Real-world scenes
- Adaptive to surveillance cameras

# 3.1 Proposed Method

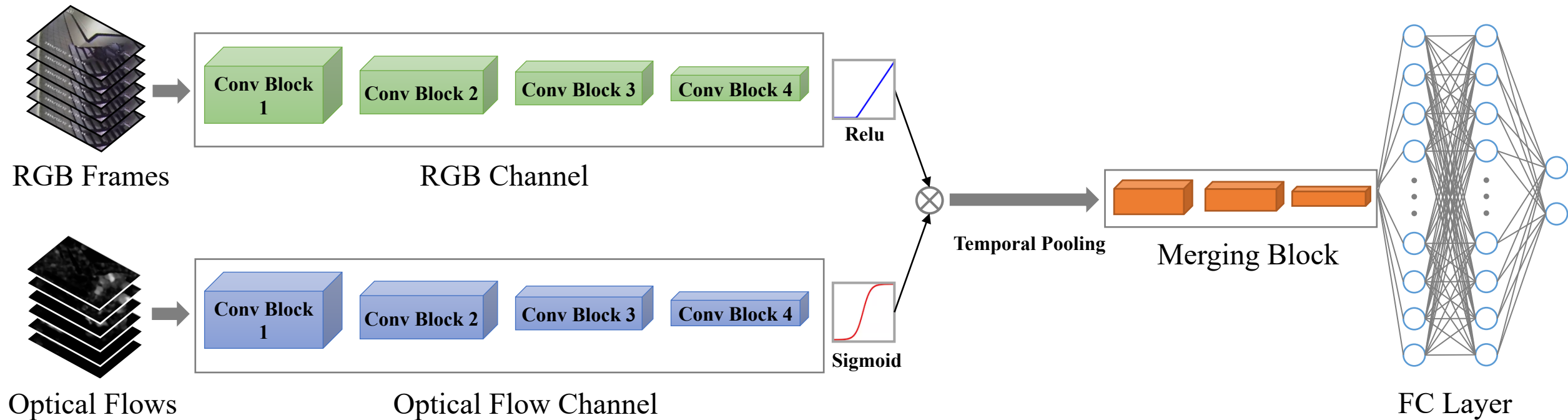


Fig 8. Structure of proposed method

# 3.2 Cropping Strategy

Optical flow is a field of 2D vector, we could calculate the norm of vector to represent magnitude of motion.

$$|\mathbf{v}(x, y)| = \sqrt{v_x^2 + v_y^2}$$

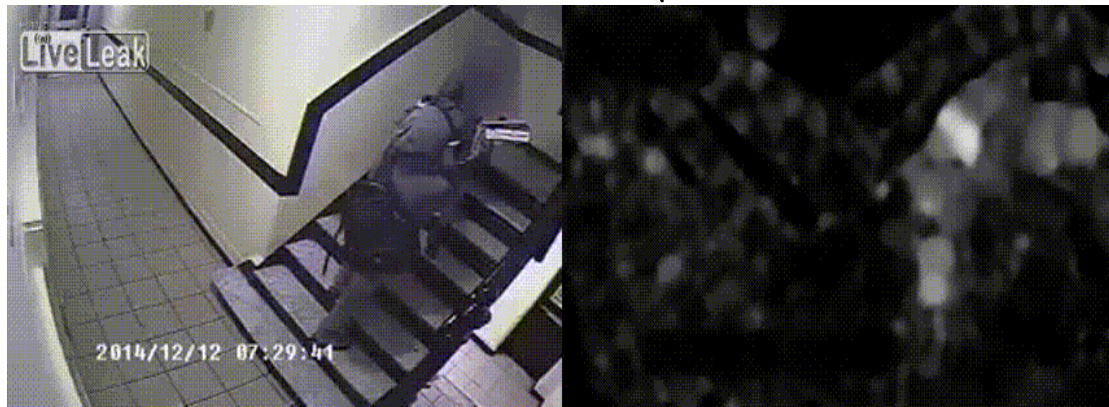


Fig 9. Motion estimation using optical flow

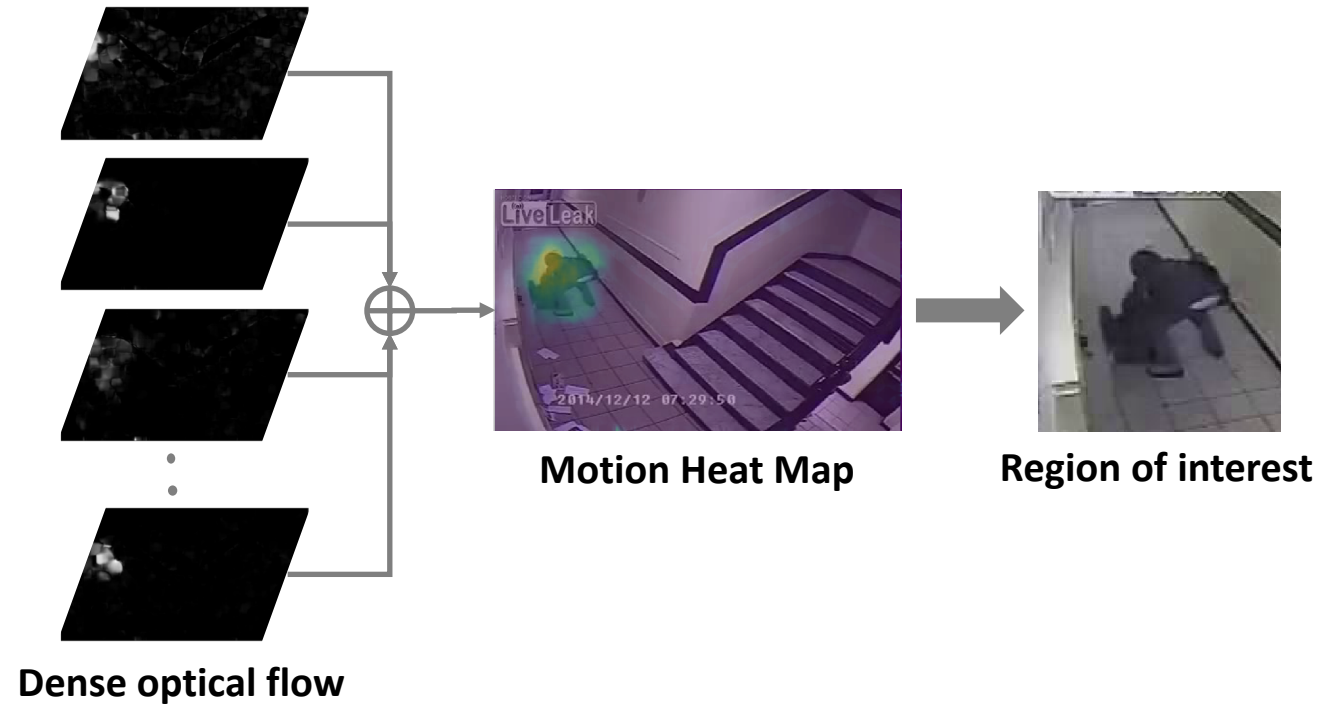


Fig 10. Extracting the region of interest

# 3.3 Sampling Strategy

Video data has much redundant information between neighboring frames, a sparse sampling strategy is implemented to reduce the amount of computing cost.

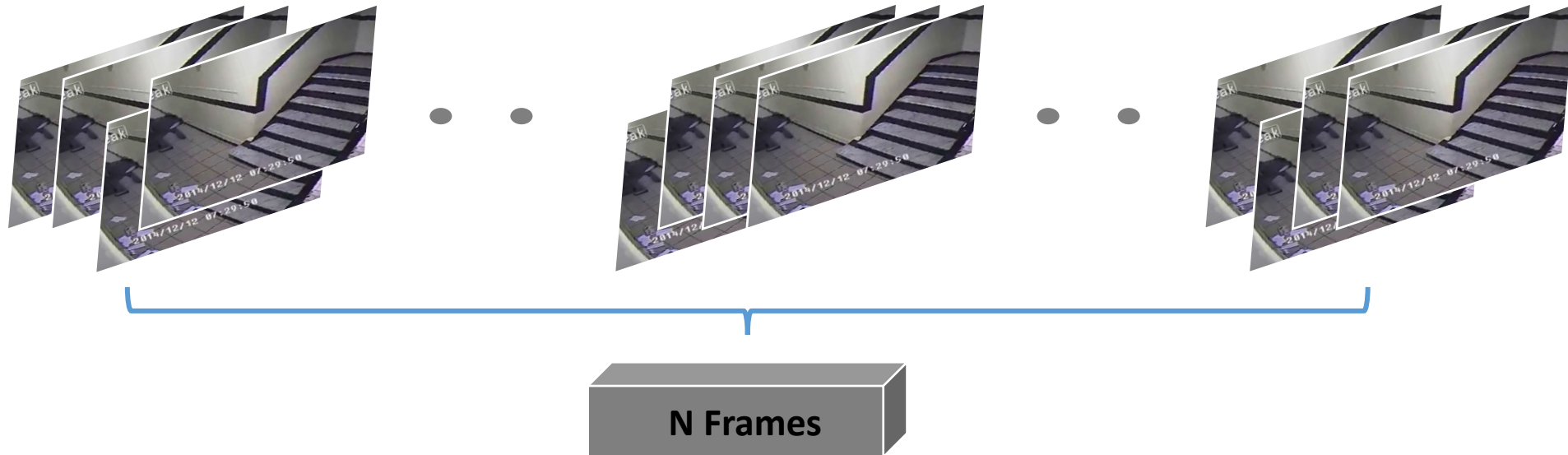


Fig 11. Sparse sampling



# 3.4 Combination

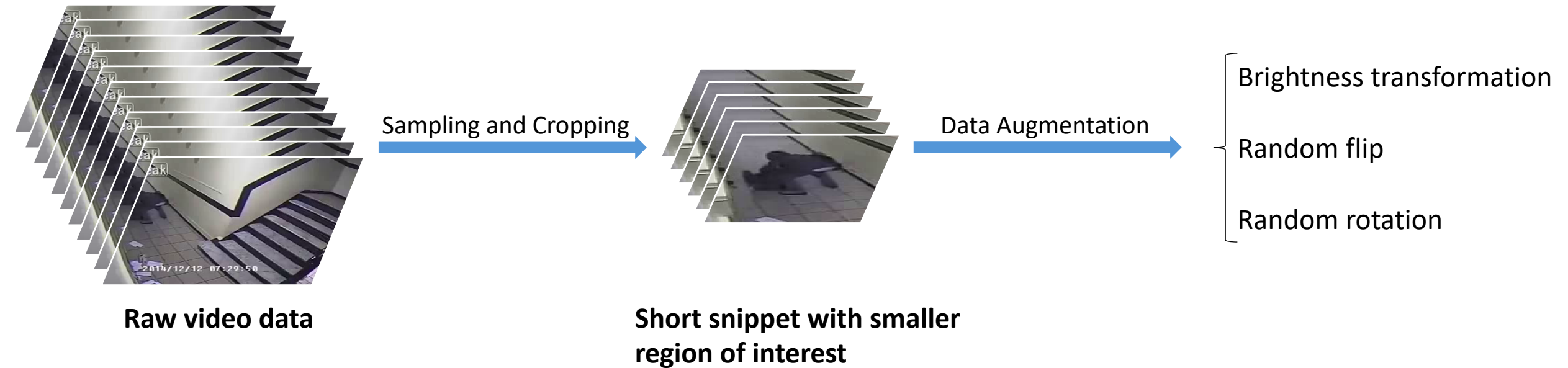


Fig 12. Combination of sampling and cropping

# 4.1 Experiments

In the training process, SGD optimizer with momentum (0.9) and learning rate decay ( $1e-6$ ) were implemented. After 6,000 iterations of training, our model obtained the best accuracy of 87.25% on the test set (shown in Table III).

Table III  
EVALUATION OF THE PROPOSED FLOW GATED NETWORK ON THE  
RWF-2000 DATASET

<b>Method</b>	<b>Train Accuracy(%)</b>	<b>Test Accuracy(%)</b>	<b>Params</b>
RGB Only	89.50	84.50	248,402
OPT Only	82.31	75.50	248,258
Fusion (P3D)	88.44	87.25	272,690
Fusion (C3D)	96.50	85.75	507,154

# 4.2 Comparisons

Table IV

COMPARISONS BETWEEN THE PROPOSED METHOD AND OTHERS ON THE PREVIOUS DATASETS

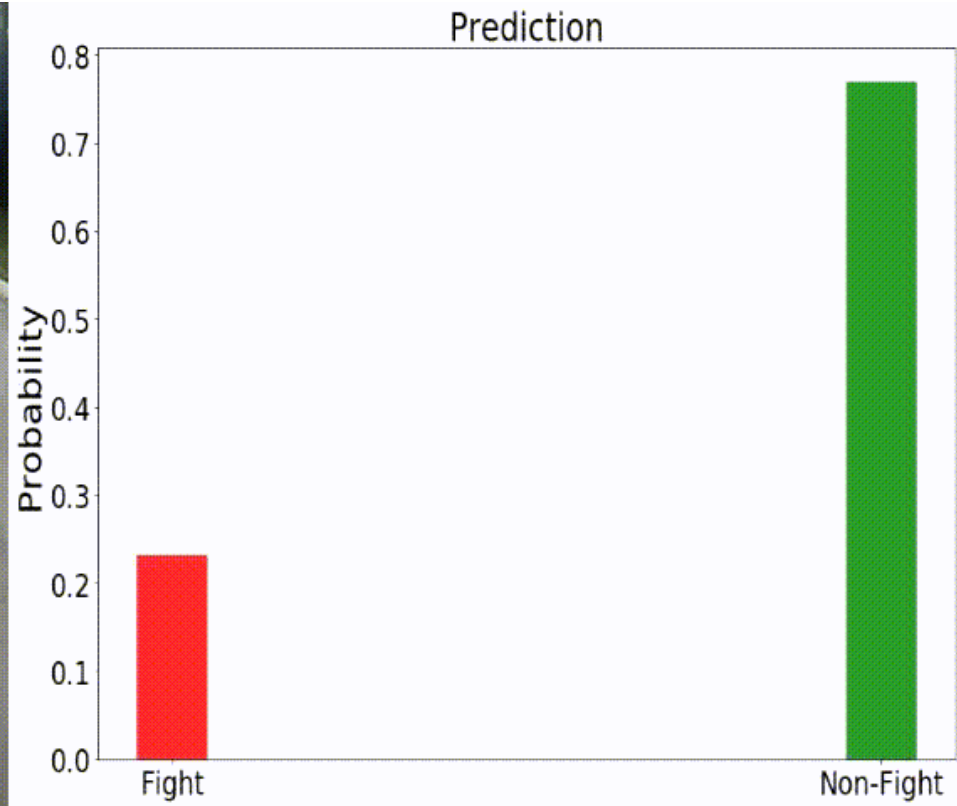
Type	Method	Movies	Hockey	Crowd
Hand-Crafted Features	ViF [6]	-	82.90%	81.30%
	LHOG+LOF [40]	-	95.10%	94.31%
	HOF+HIK [41]	59.0%	88.60%	-
	HOG+HIK [41]	49.0%	91.70%	-
	MoWLD+BoW [42]	-	91.90%	82.56%
	MoSIFT+HIK [41]	89.5%	90.90%	-
Deep-Learning Based	FightNet [26]	100%	97.00%	-
	3D ConvNet [43]	99.97%	99.62%	94.30%
	ConvLSTM [29]	100%	97.10%	94.57.
	C3D [12]	100%	96.50%	84.44%
	I3D(RGB only) [44]	100%	98.50%	86.67%
	I3D(Flow only) [44]	100%	84.00%	88.89%
	I3D(Fusion) [44]	100%	97.50%	88.89%
Ours	100%	98.00%	88.87%	

Table V

COMPARISONS BETWEEN THE PROPOSED METHOD AND OTHERS ON THE RWF-2000 DATASET

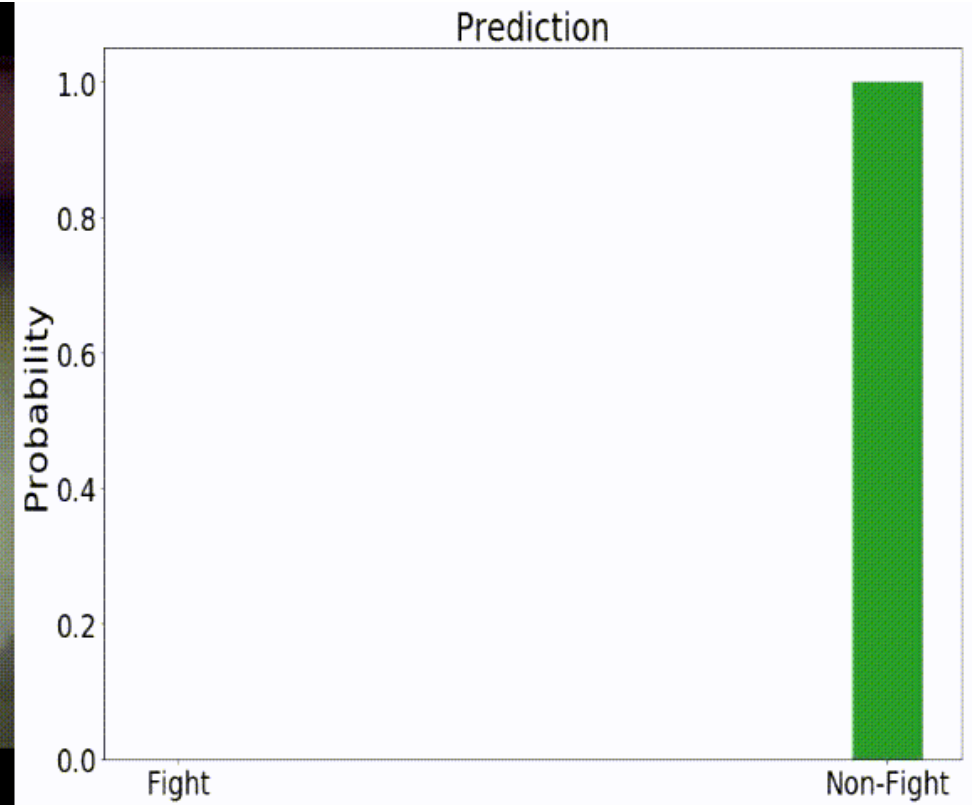
Method	Accuracy(%)	Params (M)
ConvLSTM [29]	77.00	47.4
C3D [12]	82.75	94.8
I3D (RGB only) [44]	85.75	12.3
I3D (Flow only) [44]	75.50	12.3
I3D (TwoStream) [44]	81.50	24.6
Ours (best version)	87.25	0.27

# 5.1 Demonstration





# 5.2 Demonstration



# 6 Future Work

- Videos from fixed cameras and mobile cameras could be treated differently.
- Dense optical flow is computationally expensive, an end-to-end model will be faster.
- The RWF-2000 dataset will be released as soon as possible, welcome to contact me for downloading it ([ming.cheng@dukekunshan.edu.cn](mailto:ming.cheng@dukekunshan.edu.cn)).