Temporal Segment Networks:

Towards Good Practices for Deep Action Recognition

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Motivation

- •Modeling long-range temporal structure is crucial for human activity recognition.
- •Frames in a video are highly redundant.

Modeling long-range temporal structure is not simply wrapping tons of frames.

Frames are dense, but contents are sparse!

Code Release



Temporal Segment Networks

An open source action recognition framework Winner of ActivityNet 2016 (93.2% mAP)

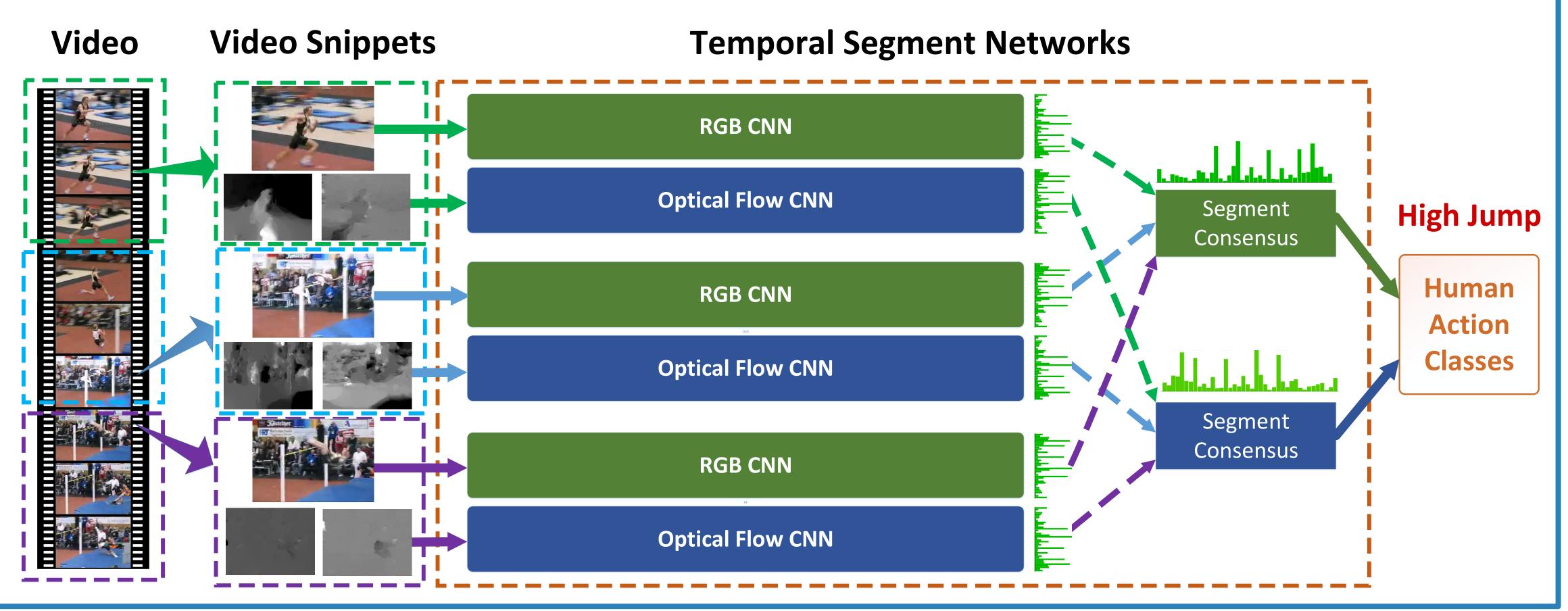
https://github.com/yjxiong/temporal-segment-networks



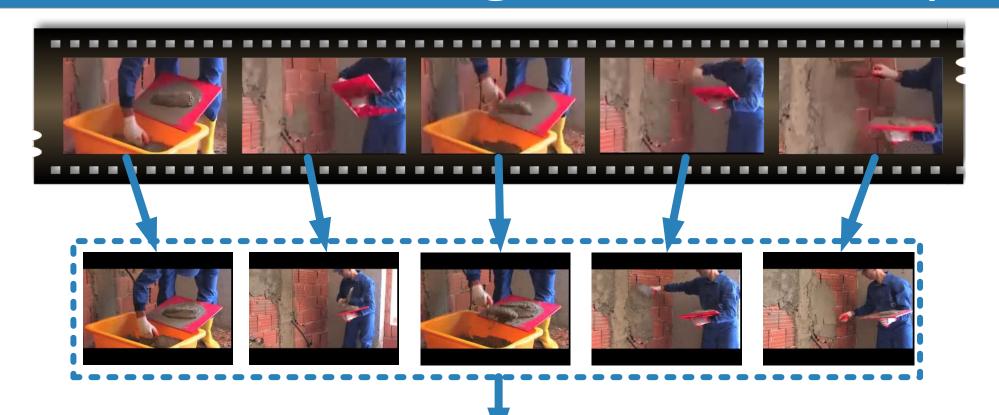
Temporal Segment Networks (TSN): The Model

Training TSN

- 1. Divide one video into a fixed number of segments
- 2. Sample snippets from the segments
- 3. Optimize the classification loss based on segment consensus



Segment-Based Sparse Sampling



- ${\cal H}$ Softmax function
- G Segment consensus function
- F ConvNet (RGB/optical flow)

 $TSN(T_1, T_2, \cdots, T_K) = \mathcal{H}(\mathcal{G}(\mathcal{F}(T_1; \mathbf{W}), \mathcal{F}(T_2; \mathbf{W}), \cdots, \mathcal{F}(T_K; \mathbf{W})))$

Segment Consensus

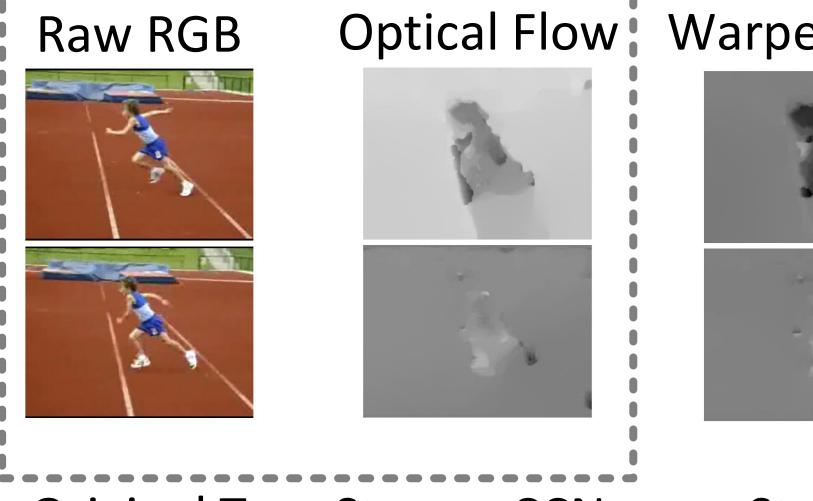
- Predict after observing all segments
- video level supervision instead of frame-wise

$$\mathcal{L}(y, \mathbf{G}) = -\sum_{i=1}^{C} y_i \left(G_i - \log \sum_{j=1}^{C} \exp G_j \right) \quad \frac{\partial \mathcal{L}(y, \mathbf{G})}{\partial \mathbf{W}} = \frac{\partial \mathcal{L}}{\partial \mathbf{G}} \sum_{k=1}^{K} \frac{\partial \mathcal{G}}{\partial \mathcal{F}(T_k)} \frac{\partial \mathcal{F}(T_k)}{\partial \mathbf{W}}$$

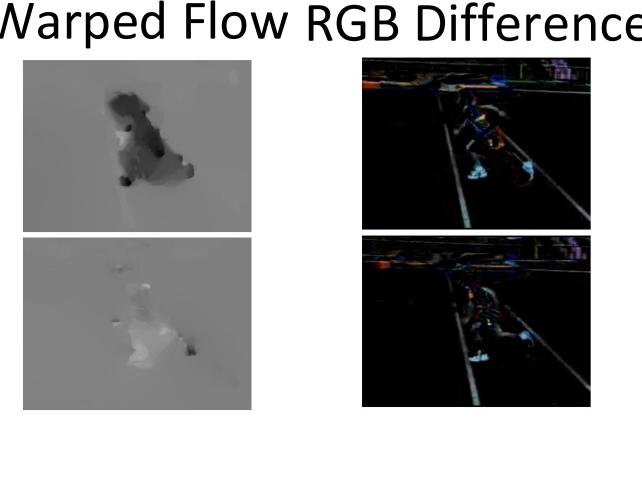
Choices for \mathcal{G}

Weighted More? Average

Input Modalities



Optical Flow: Warped Flow RGB Difference



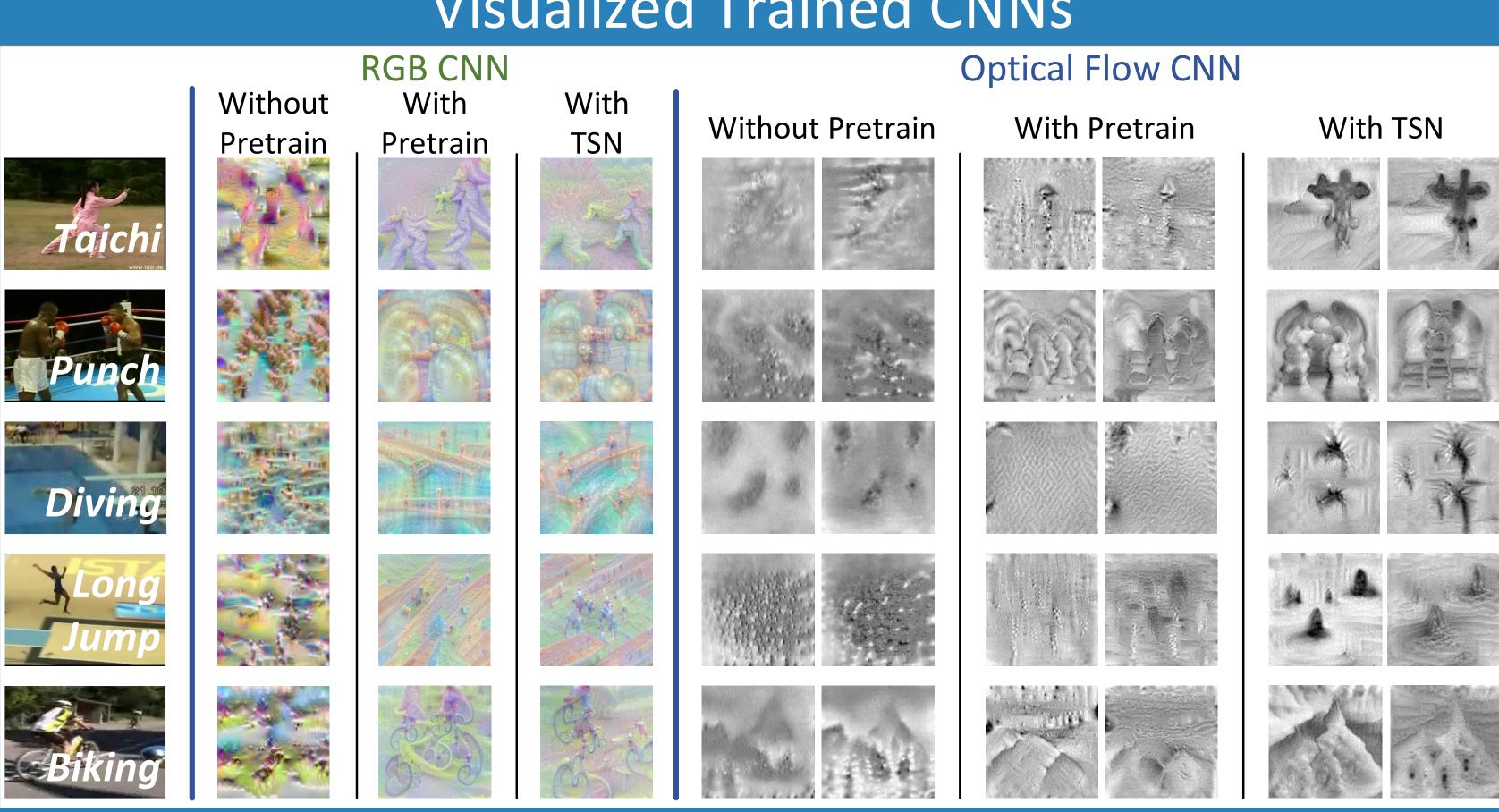
Original Two-Stream CCN

 $\Gamma SN (3 \text{ modalities})$

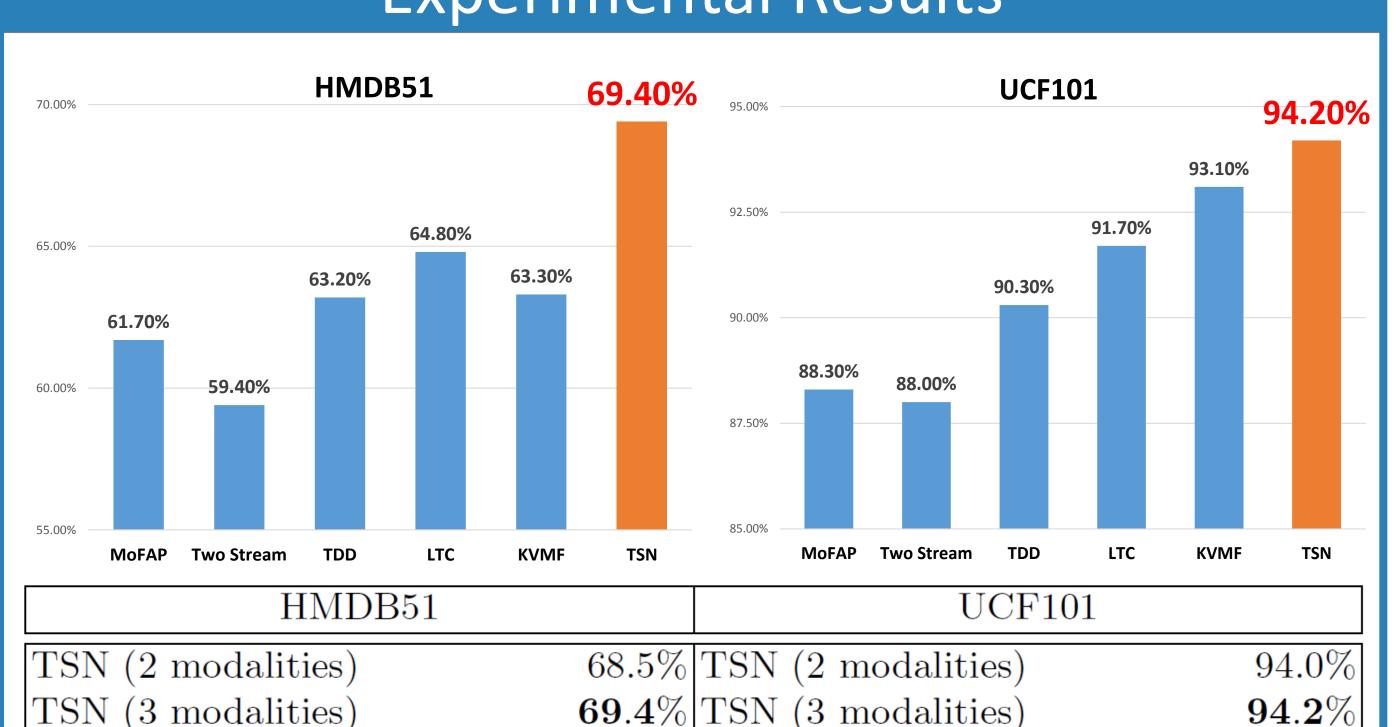
Stable Real-Time

94.2%

Visualized Trained CNNs



Experimental Results



Component Analysis: Identify the Good Practices

On UCF101-Split 1, adding components one by one

Component	Basic	Cross-Modality	Partial BN	Temporal
	Two-Stream [1]	Pre-training	with dropout	Segment Networks
Accuracy	90.0%	91.5	92.0%	93.5%

- [1] Simonyan Karen, and Andrew Zisserman. "Two-stream convolutional networks for action recognition in videos.", NIPS 2014.
- [2] Joe Yue-Hei Ng, et al. "Beyond short snippets: Deep networks for video classification." CVPR 2015.
- [3] Jeffrey Donahue, et al. "Long-term recurrent convolutional networks for visual recognition and description." CVPR. 2015.