

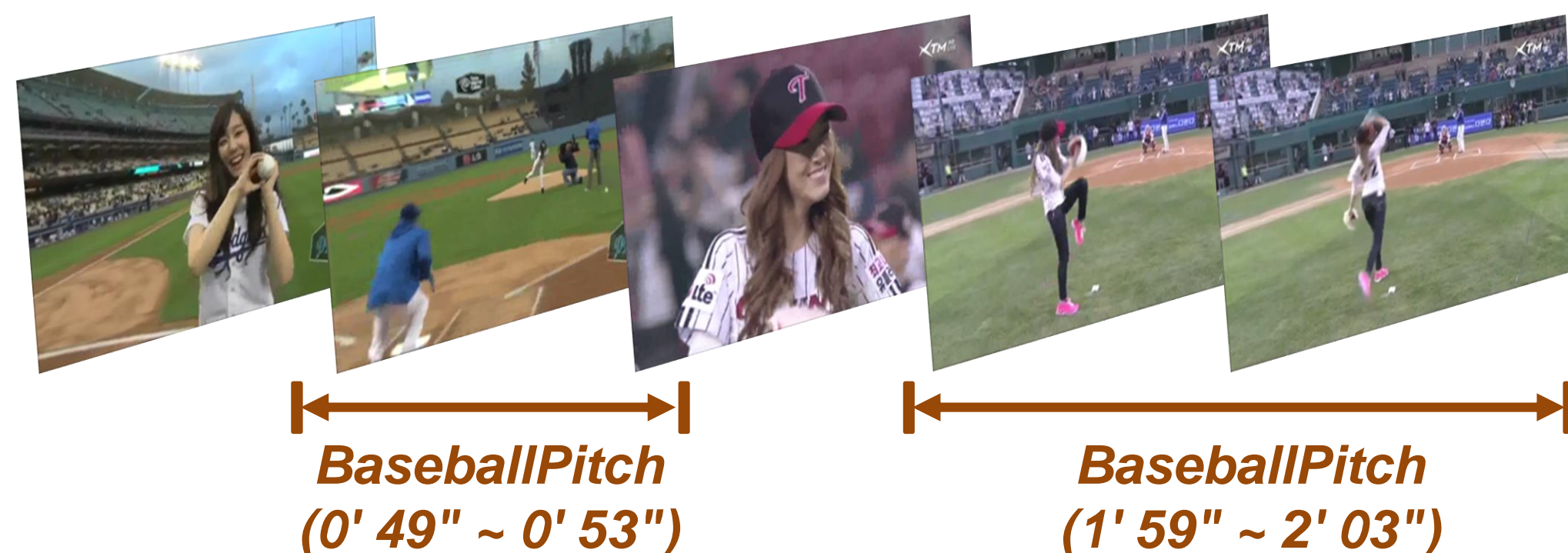
# Decomposed Cross-modal Distillation for RGB-based Temporal Action Detection

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## Problem

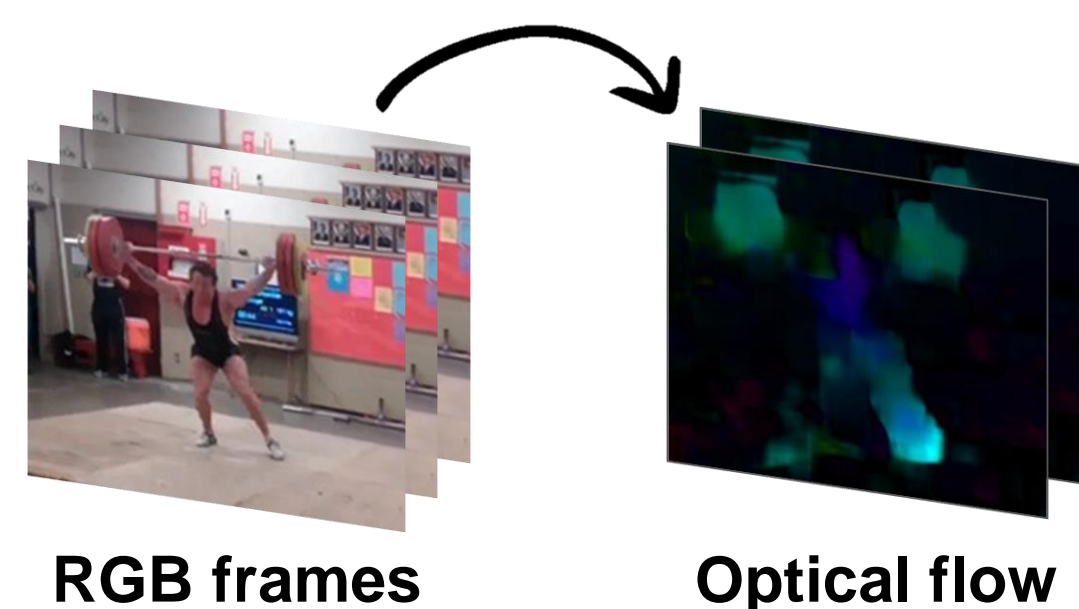
### Temporal action detection (or localization)

The goal is to predict action intervals and their classes.



## Introduction

- To date, a variety of temporal action detection models have shown promising performance based on two-stream inputs.



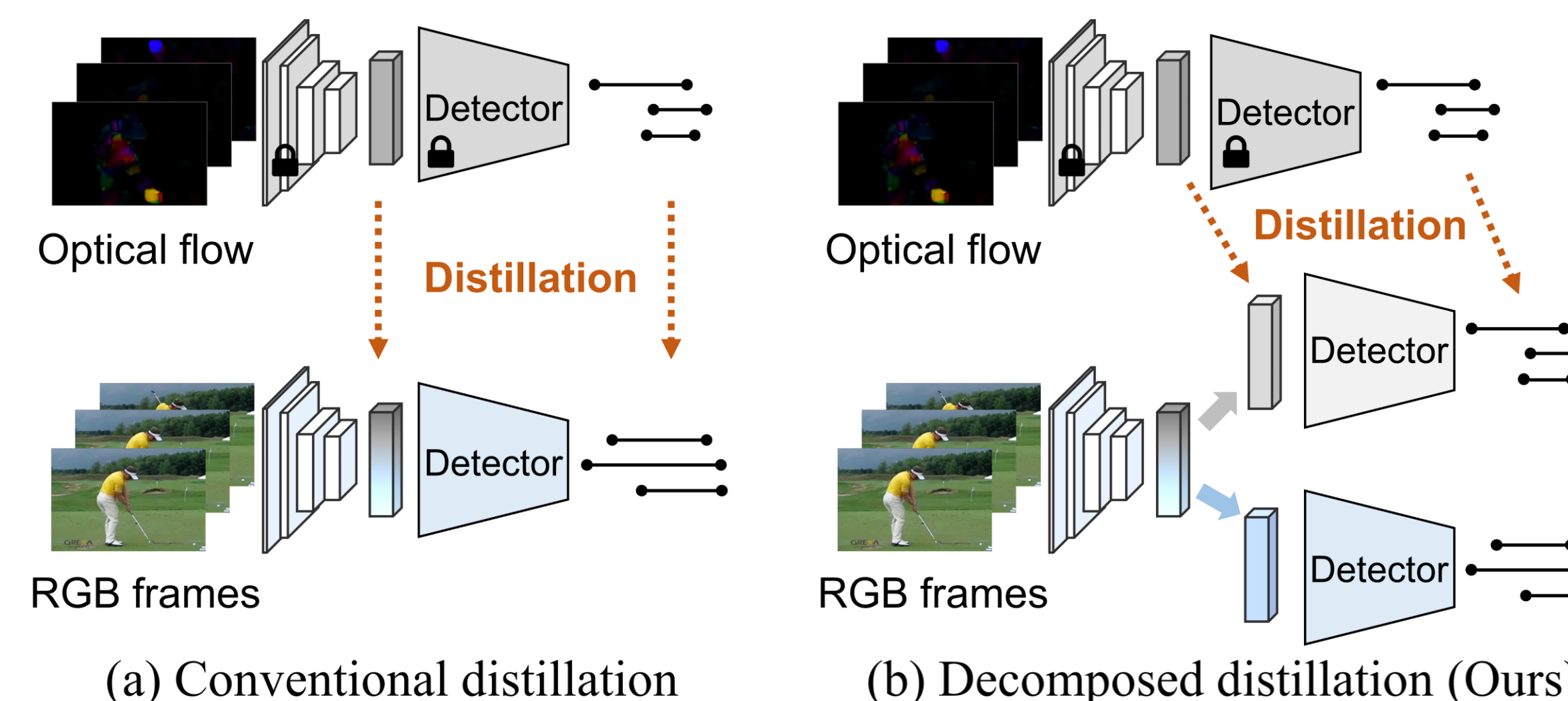
- However, they all heavily rely on computationally expensive optical flow for performance, regardless of framework types.

Framework	Method	Average mAP (%)		
		RGB+OF	RGB	$\Delta$
Anchor-based	G-TAD [74]	41.5	26.9	-14.6
	AFSD [34]	52.4	43.3	-9.1
Anchor-free	Actionformer [80]	62.2	55.5	-6.7
	TadTR [42]	56.7	46.0	-10.7
Proposal-free	TAGS [47]	52.8	47.9	-4.9

- How costly is optical flow?** E.g., TV- $L^1$  takes 1.8 minutes to process a 1-min  $224 \times 224$  video of 30 fps on a GPU.
- To bypass the cost, we aim to build a strong RGB-based action detector for both **efficient** and **accurate** prediction with a novel cross-modal knowledge distillation framework.

## Decomposed Cross-modal Distillation

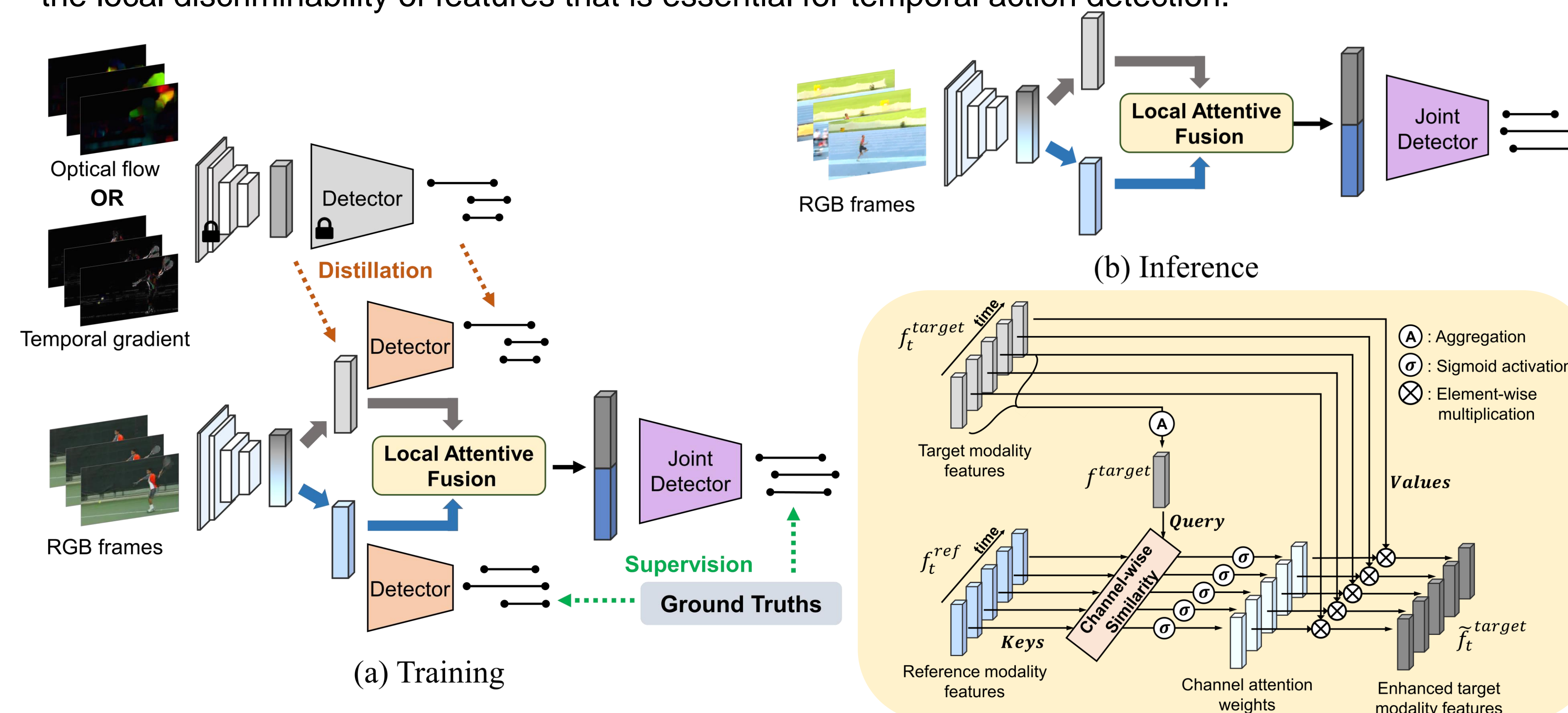
Contrary to conventional distillation where multimodal information is entangled, the proposed decomposed distillation framework encourages the model to learn it in a **decomposed way** to exploit better multimodal complementarity.



## Architecture

Our model explicitly separates the motion and appearance features within a **dual-branch pipeline** where the two branches share the detection head but pursue conflicting training objectives.

The proposed **local attentive fusion** enables effective multimodal information fusion while sustaining the local discriminability of features that is essential for temporal action detection.



## Experiments

The proposed method significantly enhances the RGB-based action detectors, while being generalizable to various backbones and detection heads.

### Ablation studies

distillation	conven.	decomp.	local attn.	mAP@IoU (%)					AVG
				0.3	0.4	0.5	0.6	0.7	
$\times$	$\times$	$\times$	$\times$	62.3	55.2	46.2	33.8	20.4	43.6
$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	62.5	55.7	47.3	35.1	21.8	44.5
				63.3	56.2	47.9	36.1	22.9	45.2
				64.4	58.0	49.0	37.5	24.1	46.6

### Generalization tests

Backbone	Distill.	mAP@IoU (%)					AVG
		0.3	0.4	0.5	0.6	0.7	
TSM18 [35]	$\times$	62.3	55.2	46.2	33.8	20.4	43.6
	TG	64.4	58.0	49.0	37.5	24.1	46.6 (+3.0)
	OF	65.3	59.5	50.9	39.6	25.5	48.2 (+4.6)
TSM50 [35]	$\times$	65.0	59.2	50.0	38.2	25.0	47.5
	TG	68.1	61.8	52.4	41.7	27.5	50.3 (+2.8)
	OF	66.5	62.3	55.3	44.5	32.9	52.3 (+4.8)
I3D [6]	$\times$	53.8	47.0	38.6	30.0	19.9	37.9
	TG	57.6	51.4	42.5	32.9	22.1	41.3 (+3.4)
	OF	57.7	52.1	44.6	34.9	24.0	42.6 (+4.7)
Slowfast50 [15]	$\times$	67.4	62.9	56.8	46.8	35.0	53.8
	TG	68.9	64.1	58.1	48.2	35.6	55.0 (+1.2)
	OF	70.5	65.8	59.2	50.1	38.2	56.8 (+3.0)

Fusion	mAP@IoU (%)					AVG
	0.3	0.4	0.5	0.6	0.7	
concat.	63.3	56.2	47.9	36.1	22.9	45.2
sum.	62.6	56.1	47.5	36.1	23.0	45.1
self-attn.	63.8	56.3	46.7	34.2	21.9	44.6
cross-attn.	63.1	54.5	46.4	35.4	21.7	44.2
diff.-attn.	61.8	54.8	46.3	32.6	21.0	43.3
local attn. (Ours)	64.4	58.0	49.0	37.5	24.1	46.6

(TG: temporal gradient, OF: Optical flow)

Head	Distill.	mAP@IoU (%)					AVG
		0.3	0.4	0.5	0.6	0.7	
G-TAD [74]	$\times$	51.4	44.7	36.0	26.4	16.8	35.1
	TG	54.8	48.9	38.1	28.0	18.1	37.6 (+2.5)
	OF	55.3	49.4	39.2	30.6	19.7	38.8 (+3.6)
TadTR [42]	$\times$	62.8	56.7	47.5	37.3	25.5	46.0
	TG	63.8	57.4	49.9	39.2	26.9	47.4 (+1.4)
	OF	64.1	58.3	51.2	40.9	28.8	48.7 (+2.7)
Actionformer [80]	$\times$	62.3	55.2	46.2	33.8	20.4	43.6
	TG	64.4	58.0	49.0	37.5	24.1	46.6 (+3.0)
	OF	65.3	59.5	50.9	39.6	25.5	48.2 (+4.6)

### State-of-the-art comparisons

Method	Venue	OF	THUMOS'14					ActivityNet1.3				
			0.3	0.4	0.5	0.6	0.7	AVG	0.5	0.75	0.95	AVG
CDC [55]	CVPR'17	$\times$	40.1	29.4	23.3	13.1	7.9	22.8	45.30	26.00	0.20	23.80
GTAN [44]	CVPR'19	$\times$	57.8	47.2	38.8	-	-	-	52.61	34.14	8.91	34.31
G-TAD* [74]	CVPR'20	$\times$	52.5	45.9	37.6	28.5	19.1	36.7	49.22	34.55	4.74	33.10
AFSD* [34]	CVPR'21	$\times$	57.7	52.8	45.4	34.9	22.0	43.6	-	-	-	32.90
TadTR* [42]	TIP'22	$\times$	59.6	54.5	47.0	37.8	26.5	45.1	49.56	35.24	9.93	34.35
E2E-TAD [40]	CVPR'22	$\times$	69.4	64.3	56.0	46.4	34.9	54.2	50.47	35.99	10.83	35.10
TAGS <sup>†</sup> [47]	ECCV'22	$\times$	59.8	57.2	50.7	42.6	29.1	47.9	54.44	34.95	8.71	34.95
Actionformer <sup>†</sup> [80]	ECCV'22	$\times$	69.8	66.0	58.7	48.3	34.6	55.5	53.21	35.15	8.03	34.94
Ours	-	$\times$	70.5	65.8	59.2	50.1	38.2	56.8	53.73	35.87	8.61	35.58

### Qualitative results

