

PointTAD: Multi-Label Temporal Action Detection with Learnable Query Points

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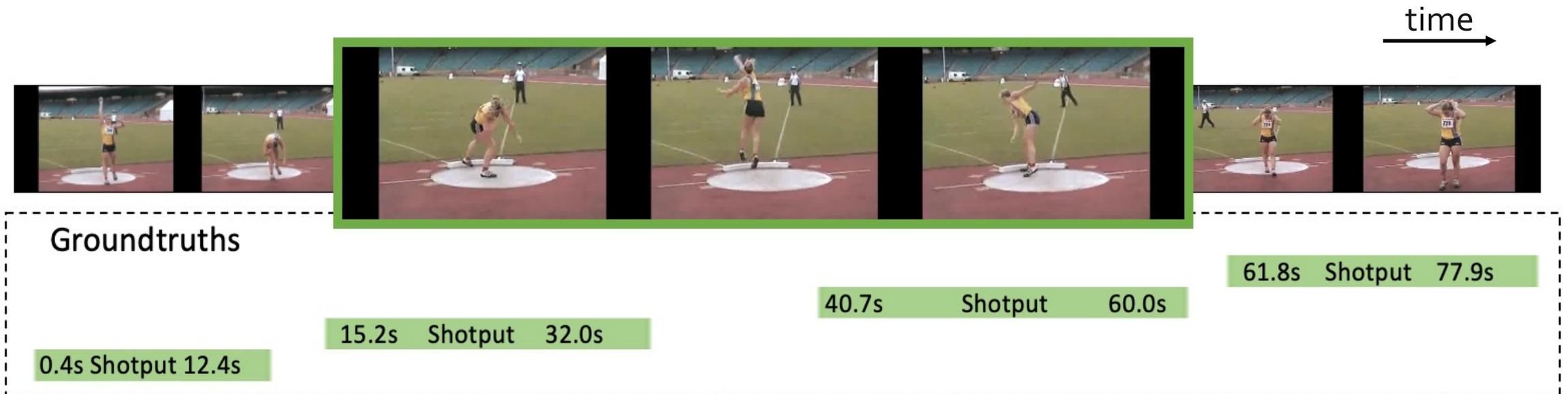
³Shanghai AI Lab



Problem

Temporal Action Detection (TAD)

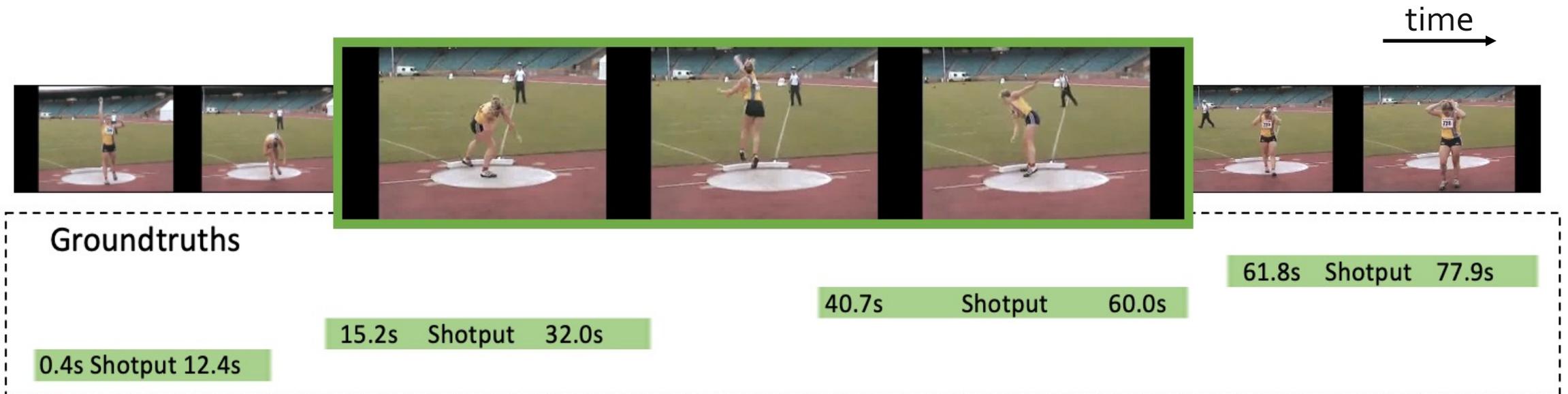
- Detecting the temporal span and class label of actions in untrimmed videos.



Problem

Temporal Action Detection (TAD)

- Non-overlapping instances
- Single-label annotations

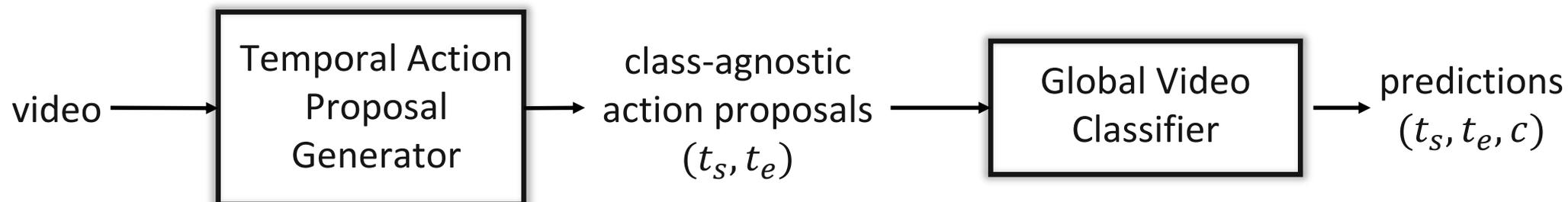


Problem

Temporal Action Detection (TAD)



Mainstream approaches:



Real-world scenario: different classes of actions often co-occur in videos!

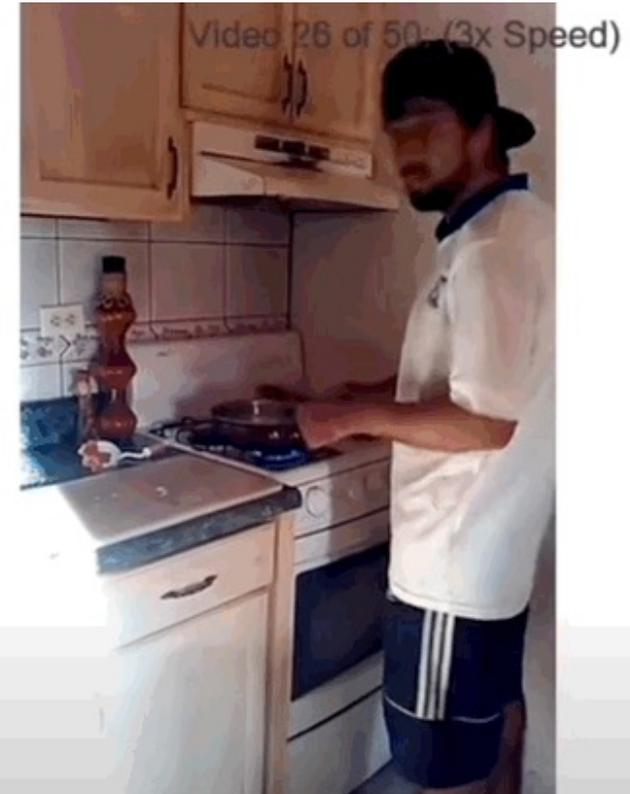
Problem

Multi-label Temporal Action Detection

- A more challenging TAD setup
 - Concurrent instances
 - Complex action relations

Annotated Actions: (gray if not active)

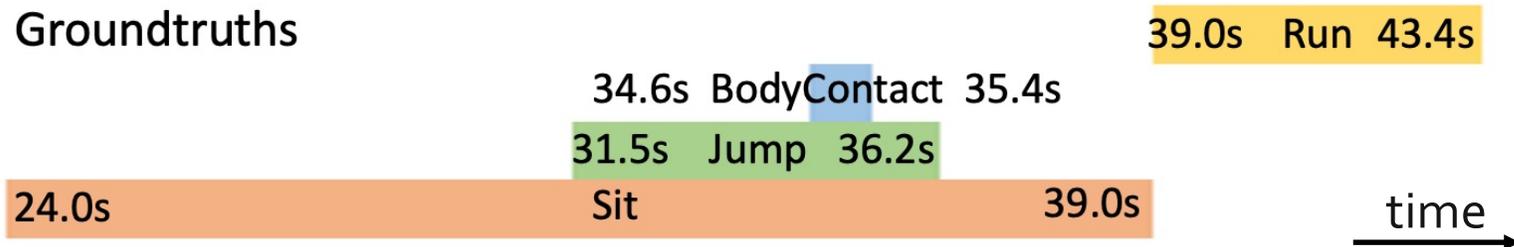
- Someone is cooking something
- Turning on a light
- Opening a refrigerator
- Taking a cup/glass/bottle from somewhere
- Closing a refrigerator
- Holding a cup/glass/bottle of something
- Drinking from a cup/glass/bottle



source: <https://prior.allenai.org/projects/charades>



Groundtruths



Related work

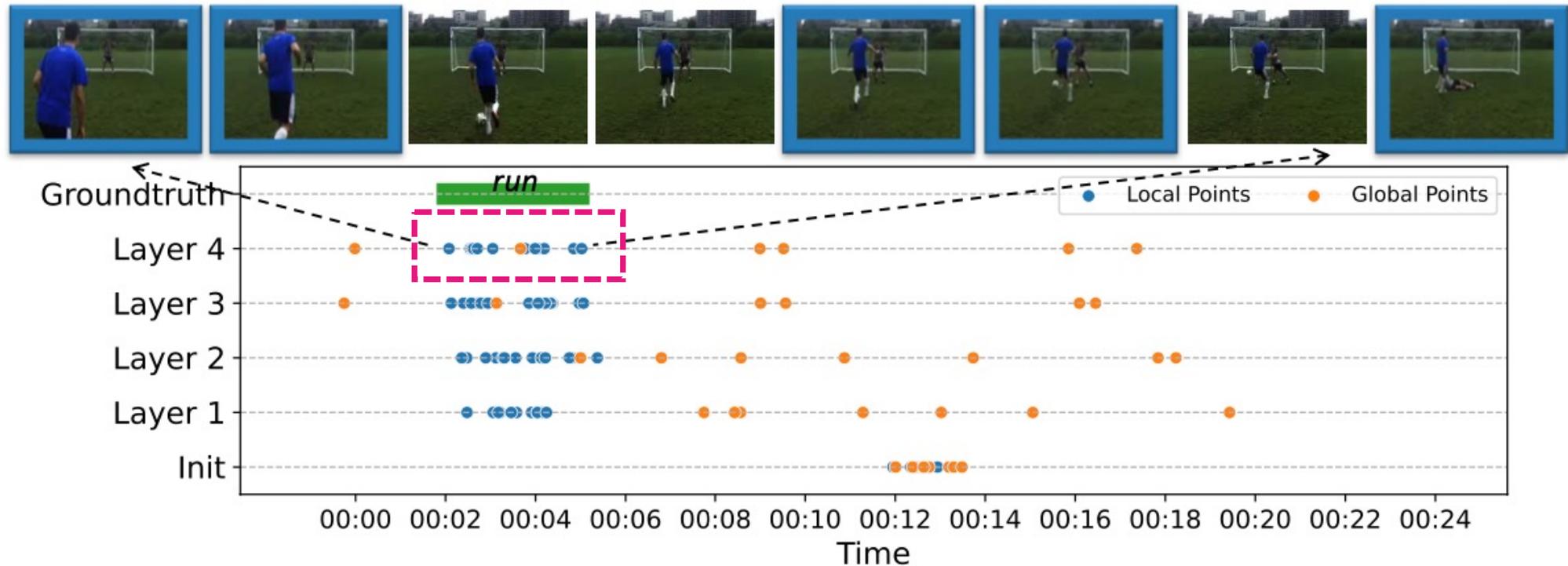
Segment-based TAD methods

- ▶ **Segment-based TAD methods** either capture incomplete action segments or get misclassified over good localization.

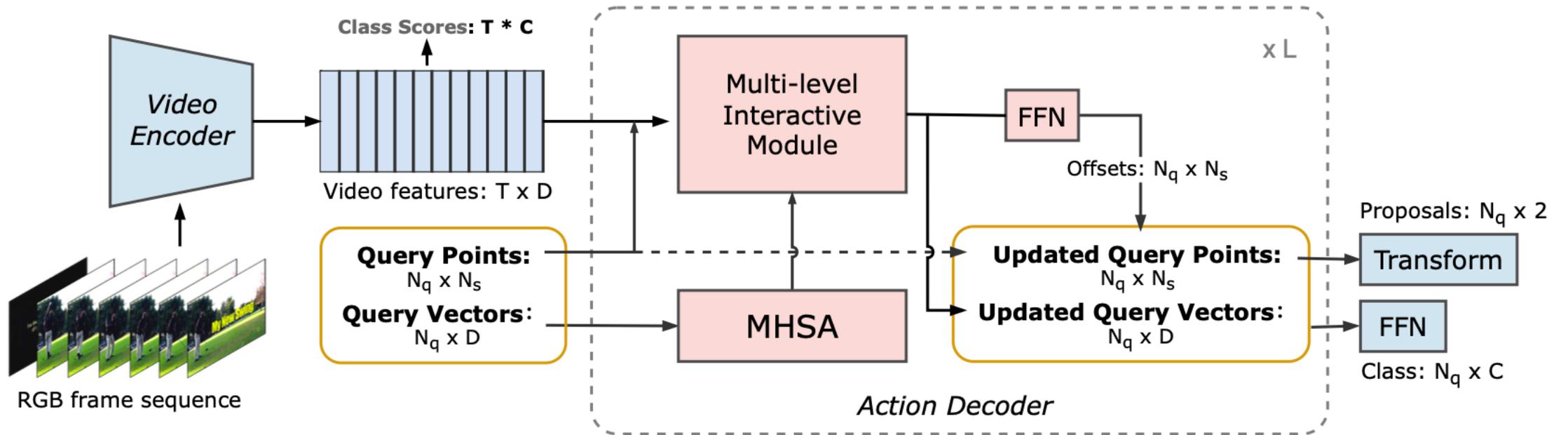


Our approach

- We introduce **Learnable Points** to handle both boundary frames and semantic keyframes of actions.

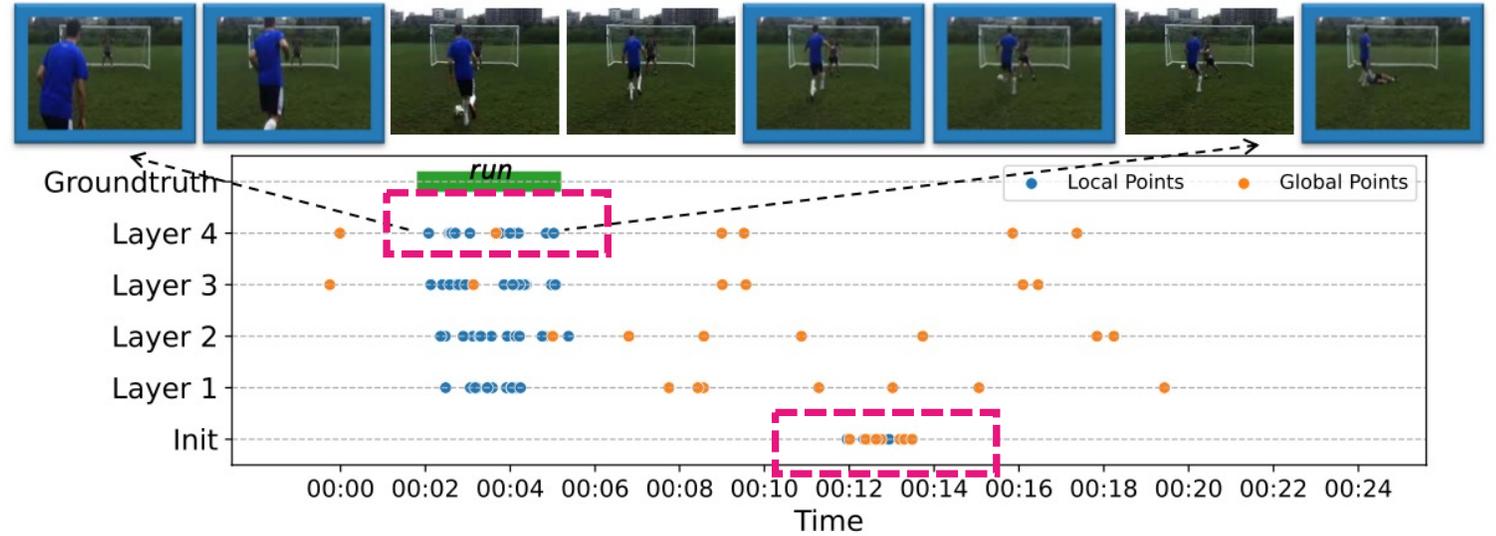


Method Overview

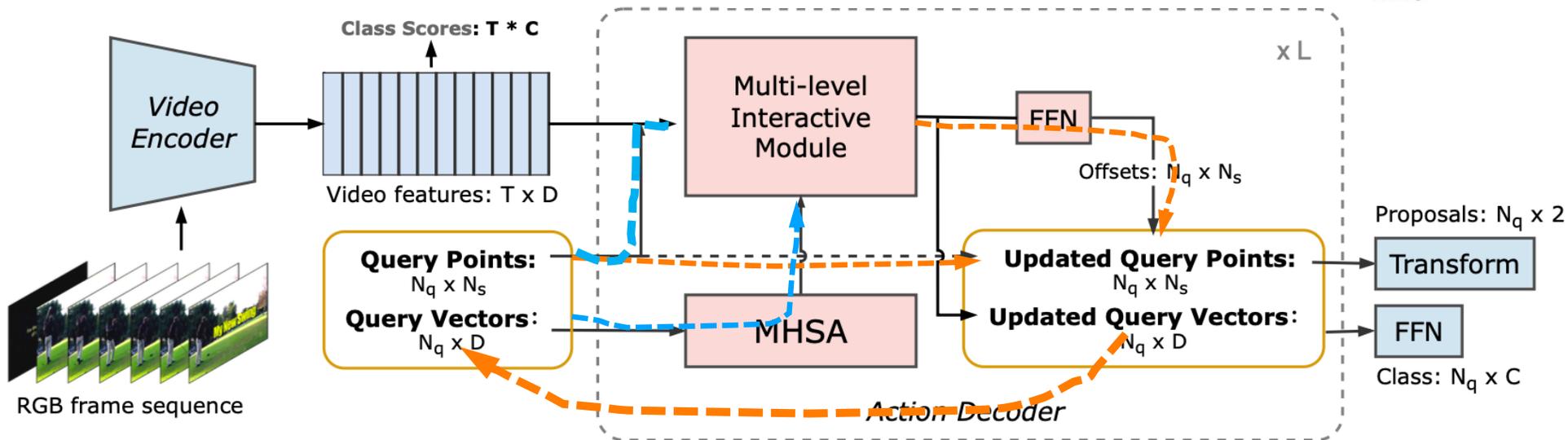


Method

Learnable Query Points



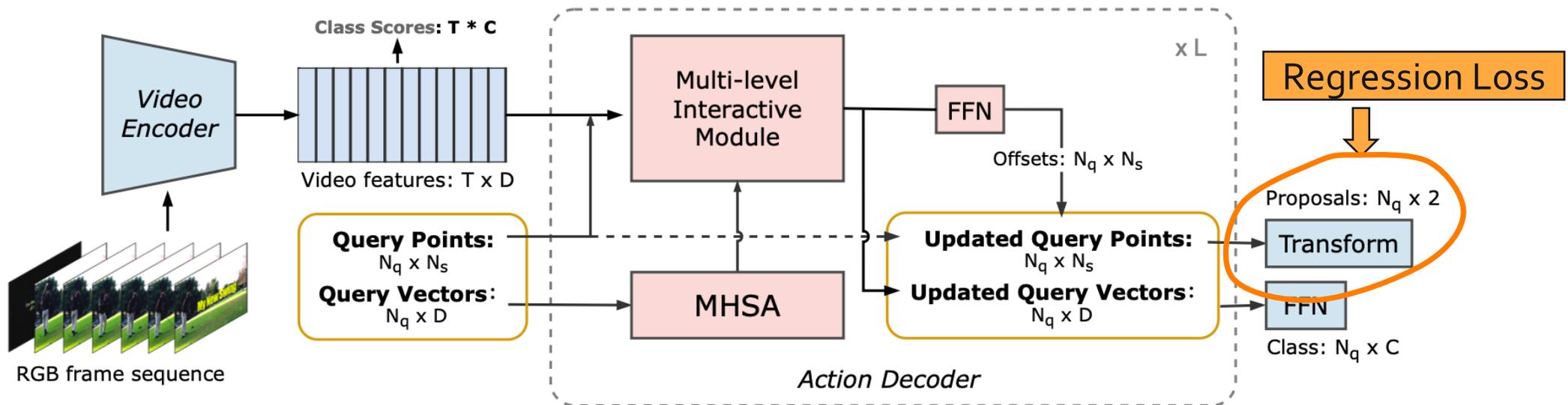
- Iterative Point Refinement.



Method

Learnable Query Points

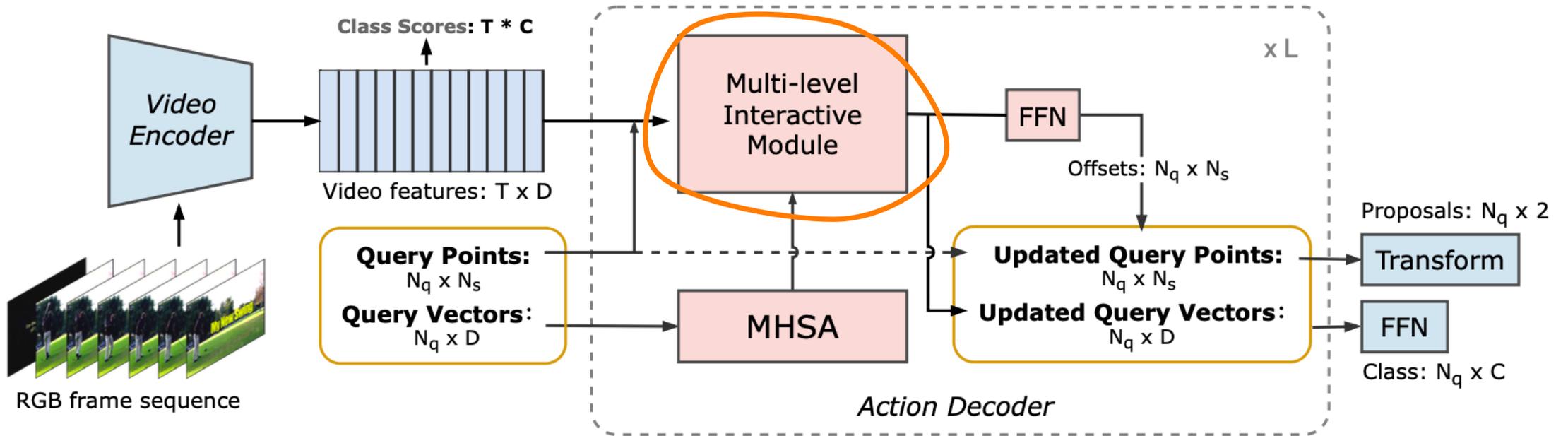
- Learning Query Points with pseudo segments
 - Point to segment transformation $\mathcal{T}: \mathcal{P} = \{t_j\}_{j=1}^{N_s} \rightarrow \mathcal{S} = (t^s, t^e)$



Method

Multi-level Interactive Module

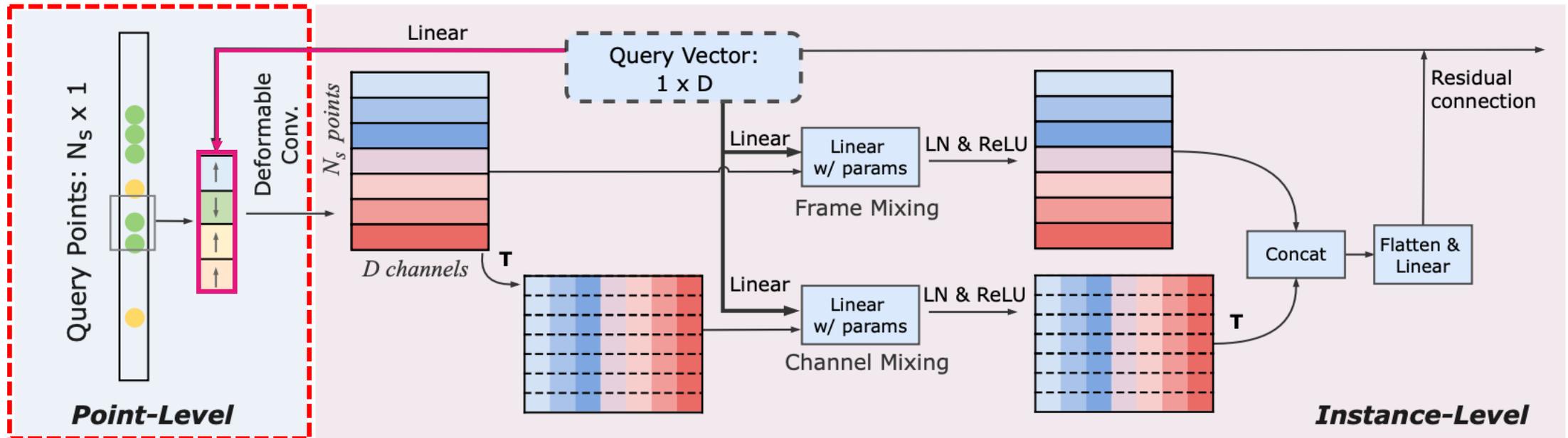
- Point-level local deformation
- Instance-level adaptive mixing



Method

Multi-level Interactive Module

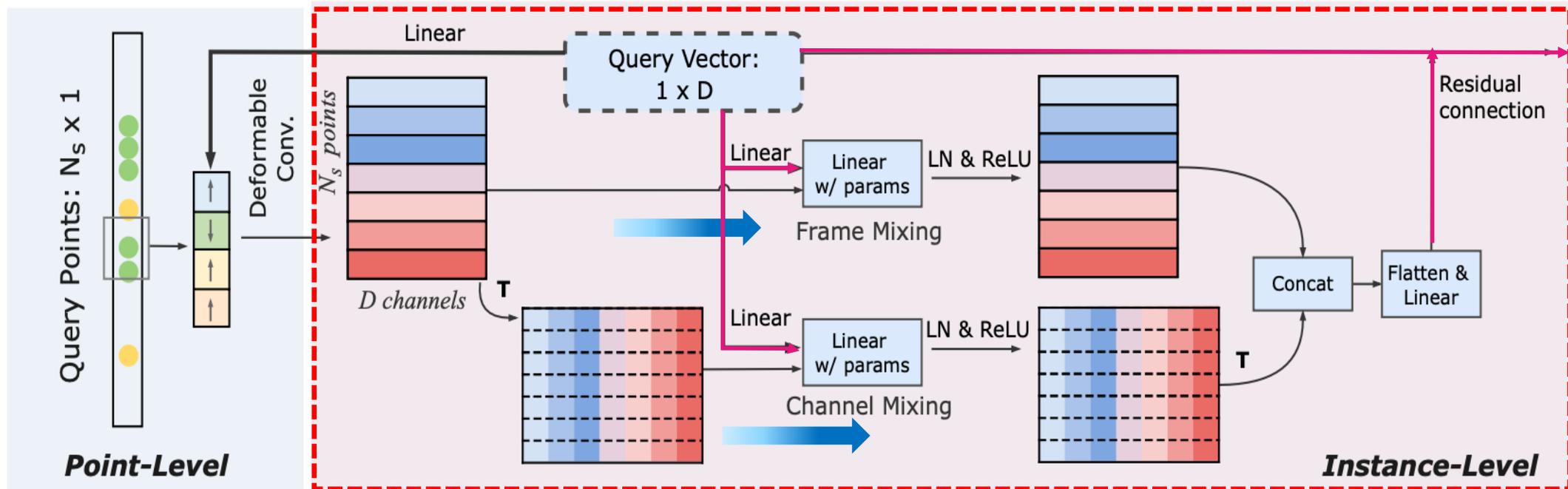
- **Point-level local deformation:** capture local temporal cue.



Method

Multi-level Interactive Module

- **Instance-level adaptive mixing:** explore frame relations and channel dynamics
 - **Channel-wise** and **frame-wise** dynamic convolution in **parallel**



Results

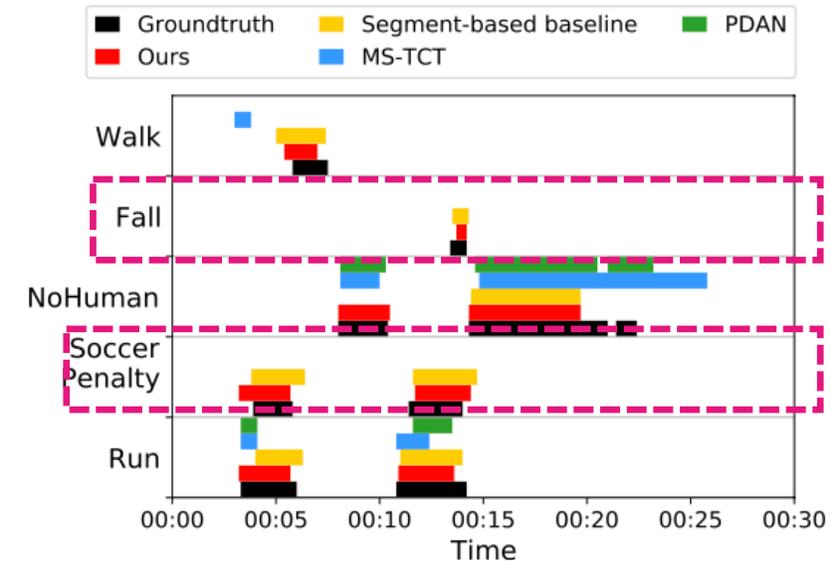
- Quantitative comparison
 - Introduced detection-mAP from classic TAD for instance-wise detection evaluation.

Table 1: Comparison with the state of the art on the MultiTHUMOS test set and Charades test set, under detection-mAP (%) and segmentation-mAP(%).

Methods	Modality	MultiTHUMOS		Charades	
		Det-mAP	Seg-mAP	Det-mAP	Seg-mAP
R-C3D [44]	RGB	-	-	-	17.6
Super-event [29]	RGB	-	36.4	-	18.6
TGM [30]	RGB	-	37.2	-	20.6
TGM [30]	RGB+OF	-	44.3	-	21.5
PDAN [9]	RGB	17.3/17.1*	40.2	8.5	23.7
Coarse-Fine [16]	RGB	-	-	6.1	25.1
MLAD [40]	RGB	14.2/13.9*	42.2	-	18.4
MLAD [40]	RGB+OF	-	51.5	-	23.7
CTRN [7]	RGB	-	44.0	-	25.3
CTRN [7]	RGB+OF	-	51.2	-	27.8
AGT [27]	RGB+OF	-	-	-	28.6
MS-TCT [8]	RGB	16.2/16.0*	43.1	7.9	25.4
Ours	RGB	21.5/21.4*	39.8	11.1	21.0

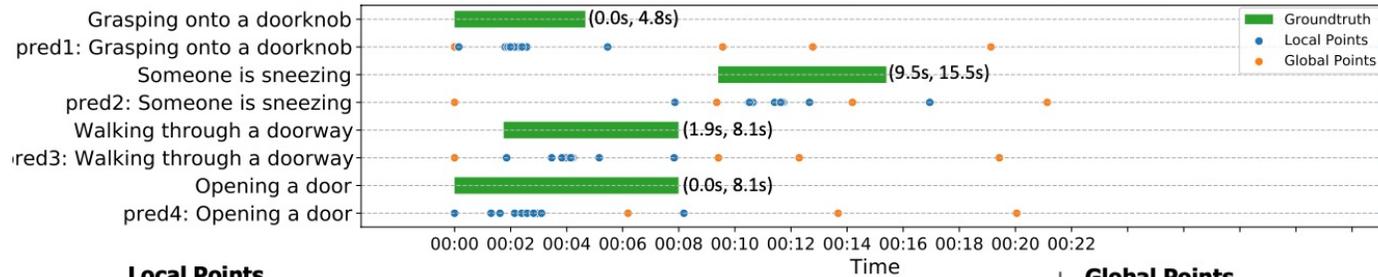
* indicates detection results excluding NoHuman class.

- Qualitative comparison

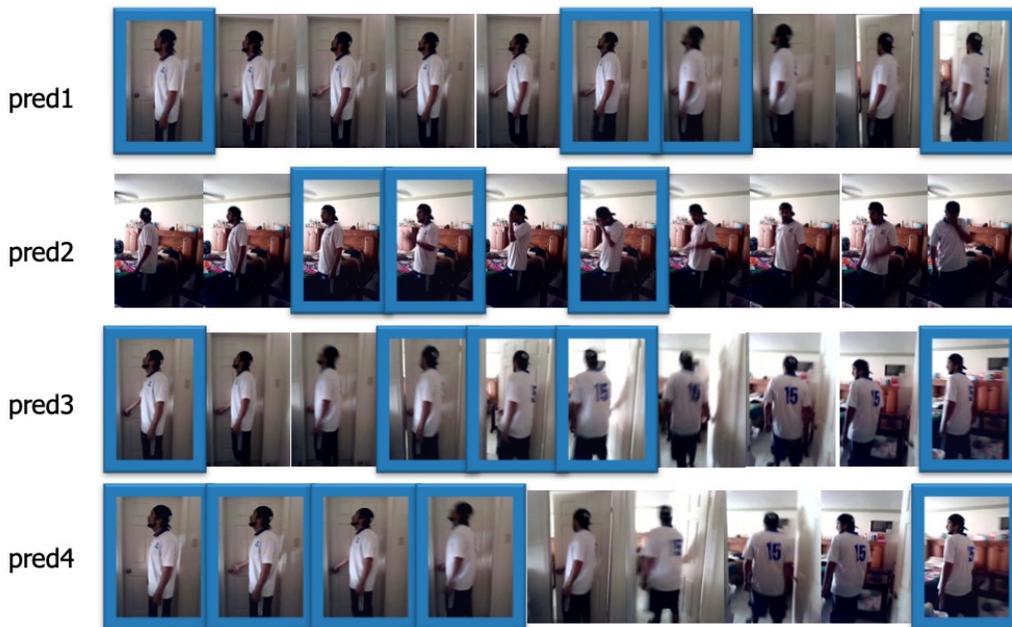


- More precise temporal boundaries
- More instances detected at harder categories

Visualization



Local Points



Global Points



- For highly overlapping groundtruths, local query points capture different frames for different actions.

Thanks for your attention!

Code is available at <https://github.com/MCG-NJU/PointTAD>

