



中国传媒大学
COMMUNICATION UNIVERSITY OF CHINA



媒体融合与传播国家重点实验室
State Key Laboratory of Media Convergence and Communication

Locality-aware Attention Network with Discriminative Dynamics Learning For Weakly Supervised Anomaly Detection

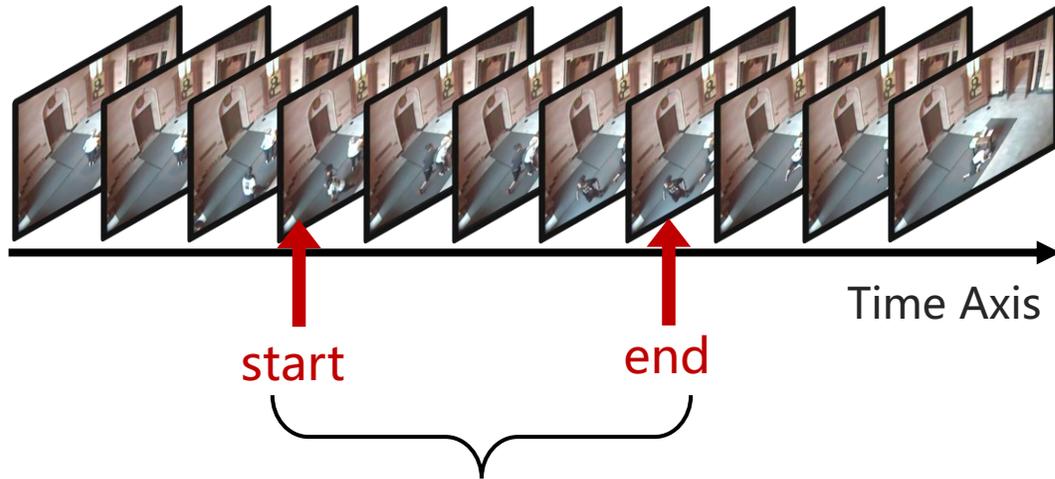
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ICME 2022

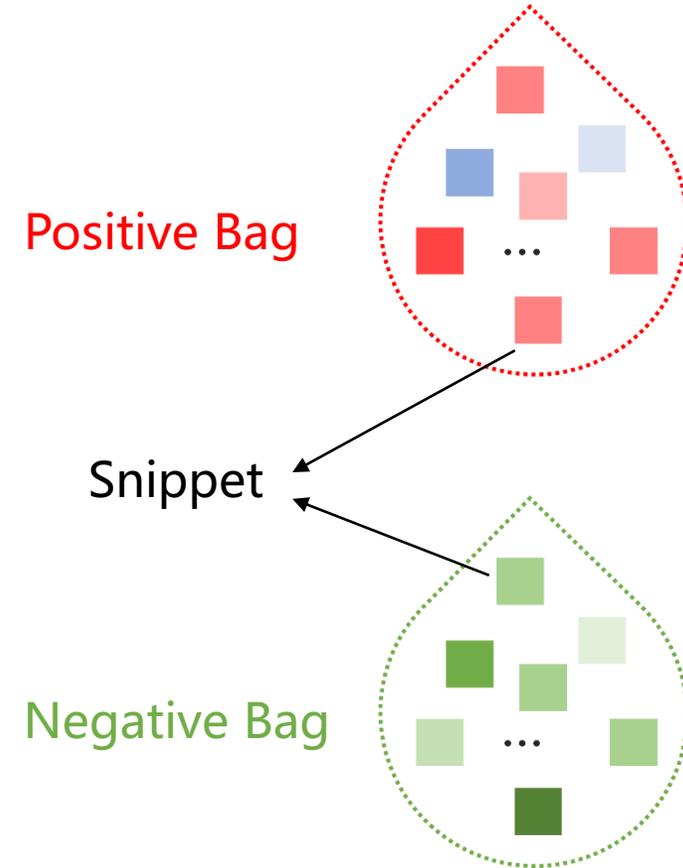
¹State Key Laboratory of Media Convergence and Communication,
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Introduction

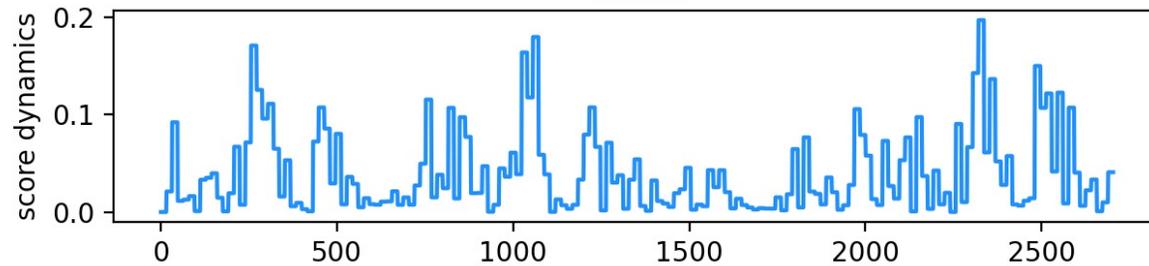
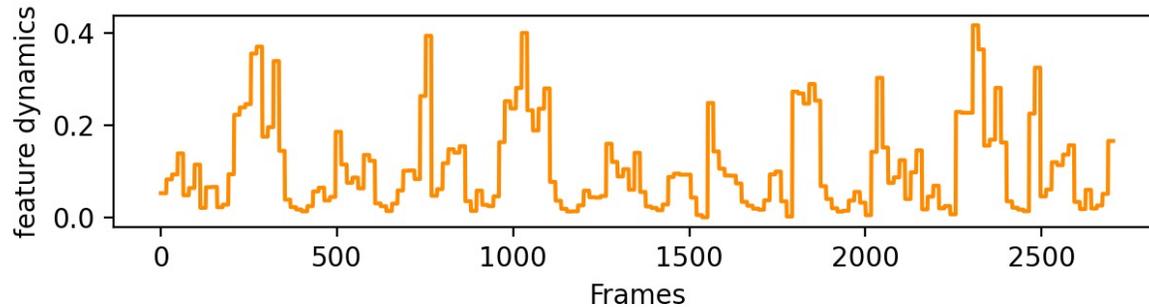
Video Anomaly Detection



locating the **start** and **end** of the event at the frame level



Motivation



Temporal Dynamics

- **Feature Dynamics (FD)**

Feature Difference between adjacent snippet

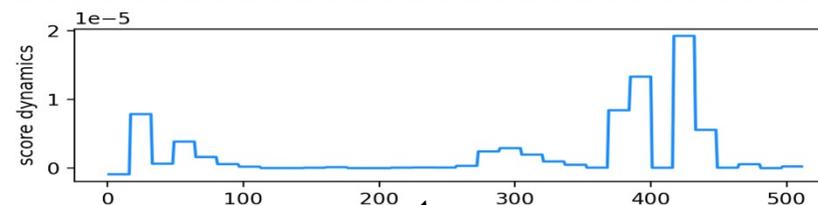
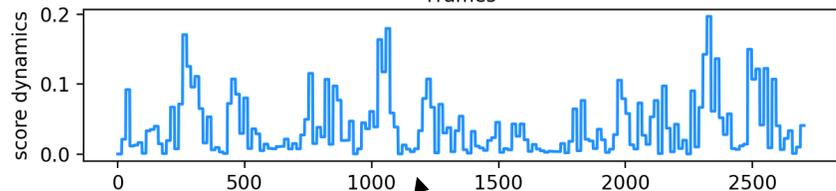
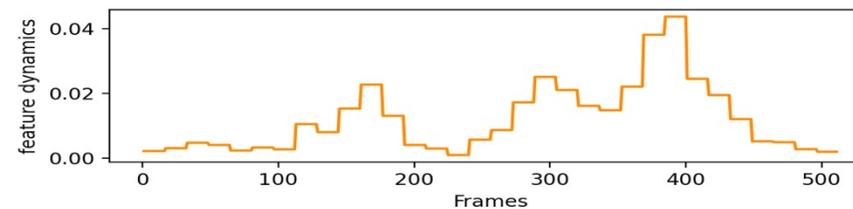
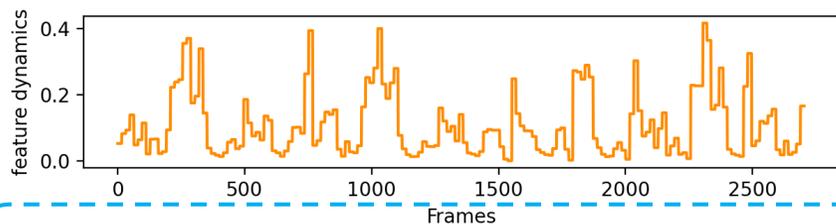
- **Score Dynamics (SD)**

Score Difference between adjacent snippet

Motivation

Positive Bag

Negative Bag



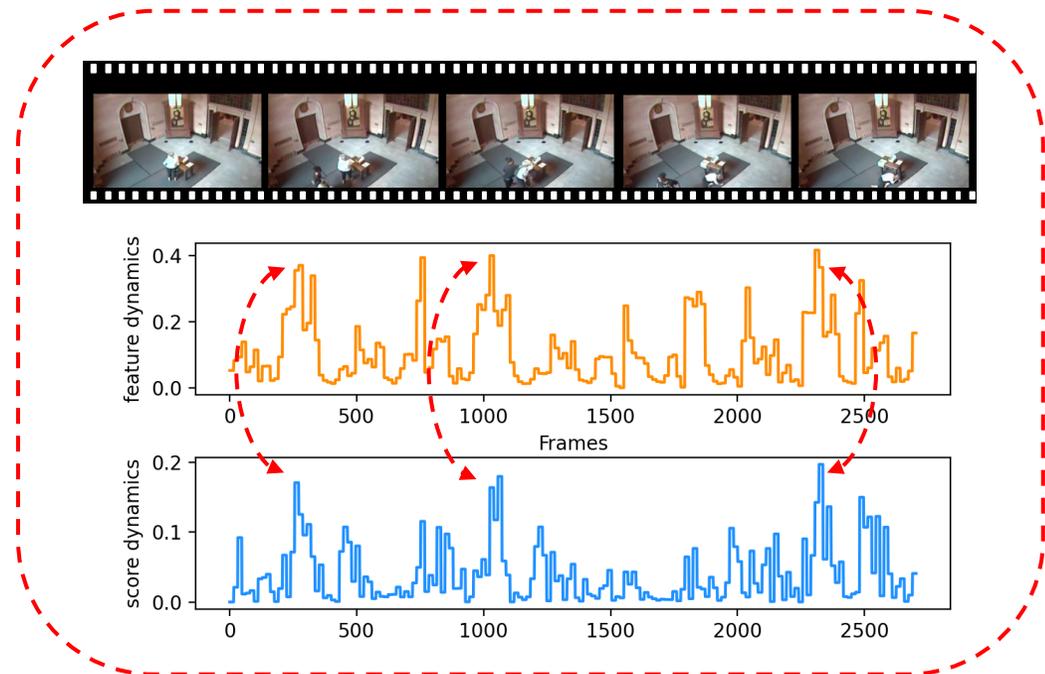
SD in Positive Bag \gg **SD** in Negative Bag



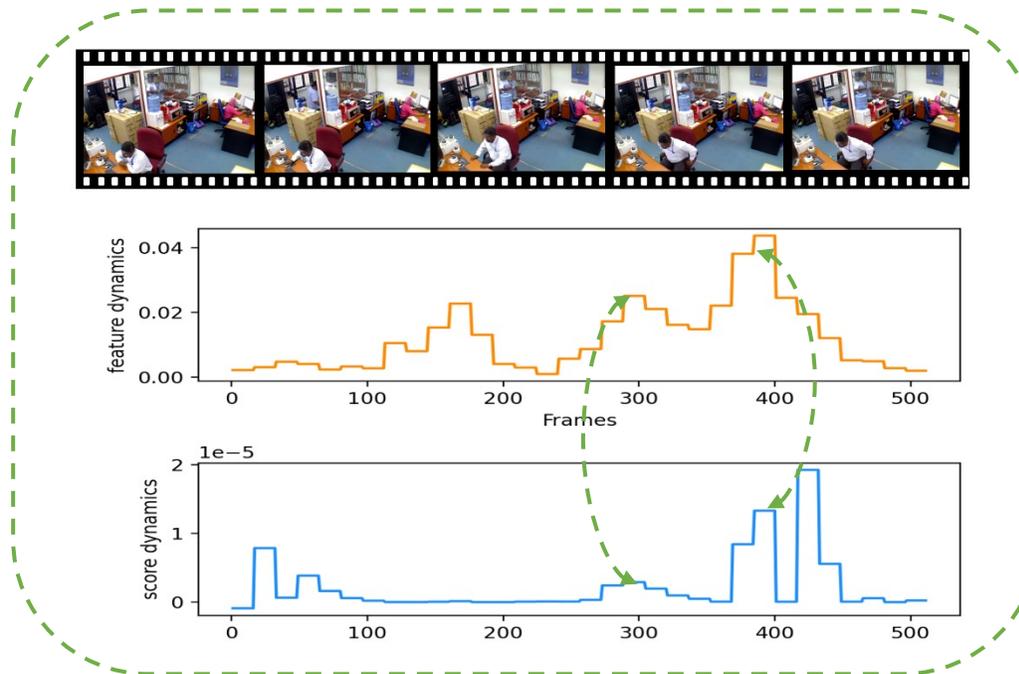
Score Dynamics Ranking \mathcal{L}_{DR}

Motivation

Positive Bag



Negative Bag



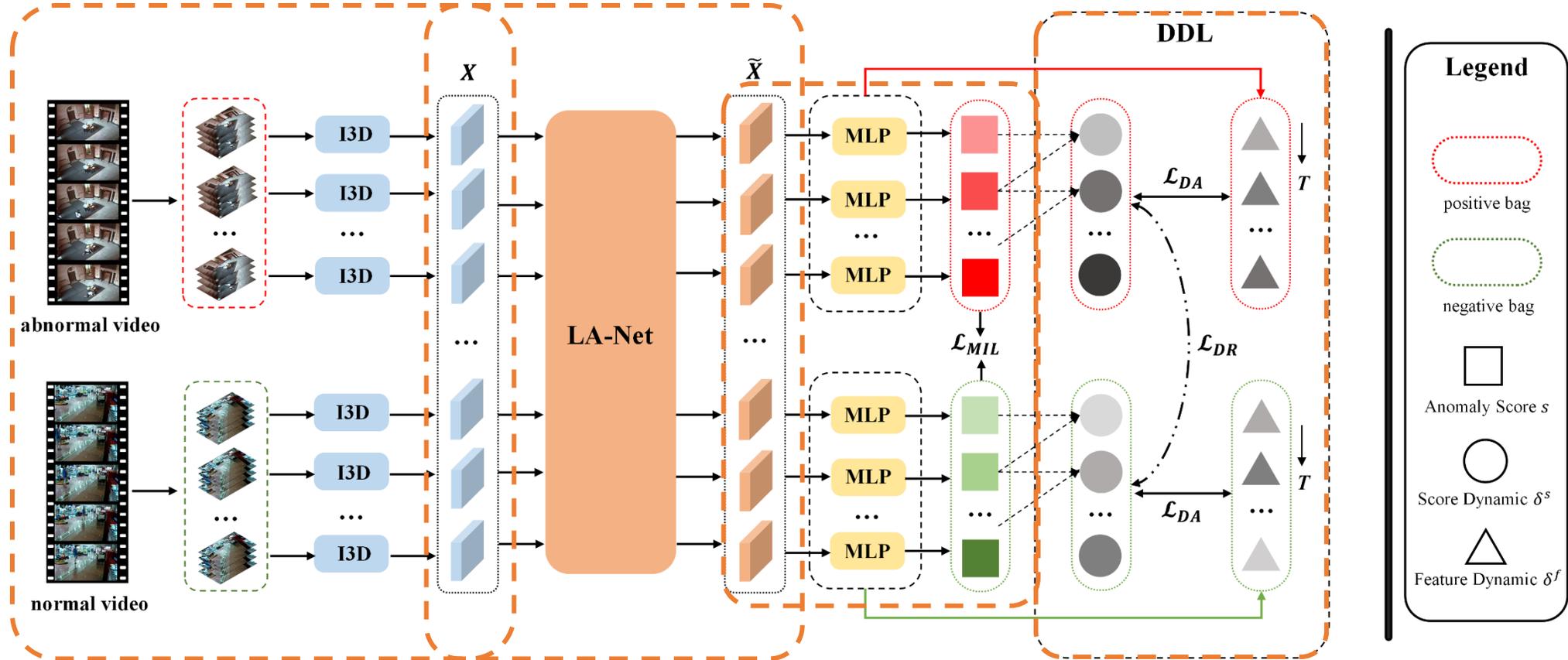
The **FD** and **SD** within a bag show a certain temporal consistency



Feature Dynamics Alignment \mathcal{L}_{DA}

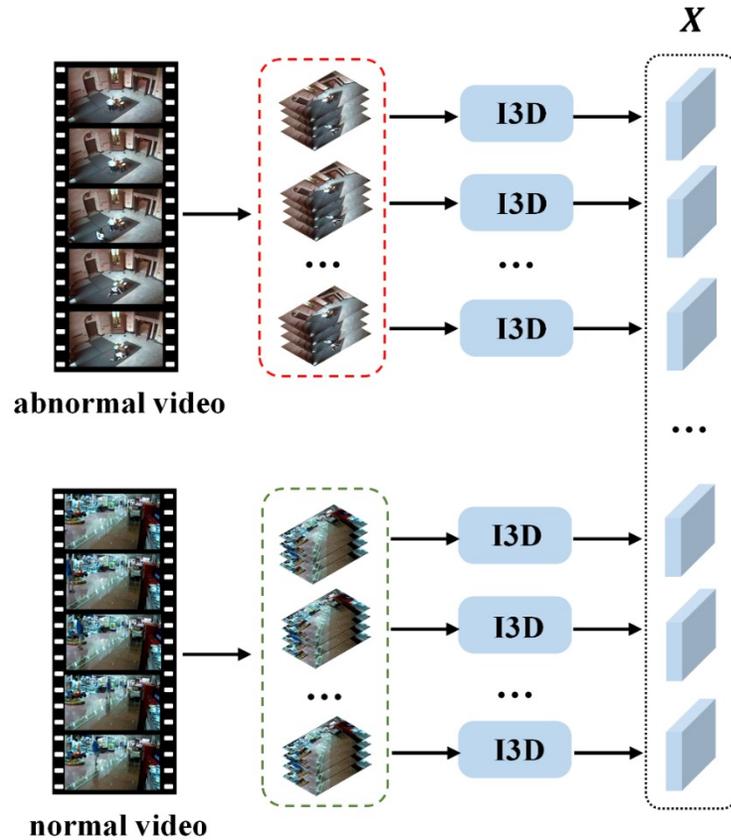
Methodology

Overall Structure of LA-Net with DDL method



Methodology

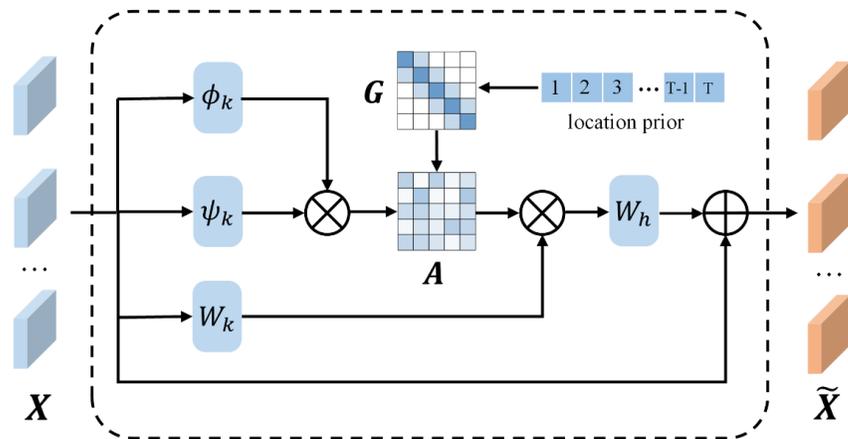
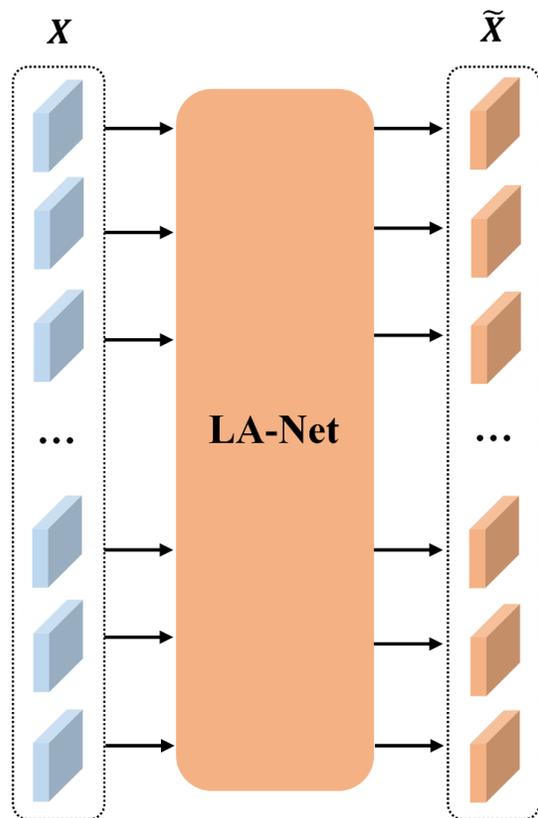
Feature Extraction



Untrimmed Video \mathcal{V}
Sliding Window $\chi = \{x_i\}_{i=1}^T$
Snippet Feature $X = I3D(\chi) \in \mathbb{R}^{T \times D}$

Methodology

Locality-aware Attention Network



$$A_{ij} = \frac{\exp\{\mathcal{R}_k(x_i, x_j)\}}{\sum_{n=1}^T \exp\{\mathcal{R}_k(x_i, x_n)\}}$$

$$\mathcal{R}_k = \phi(x_i)^T \psi(x_j)$$

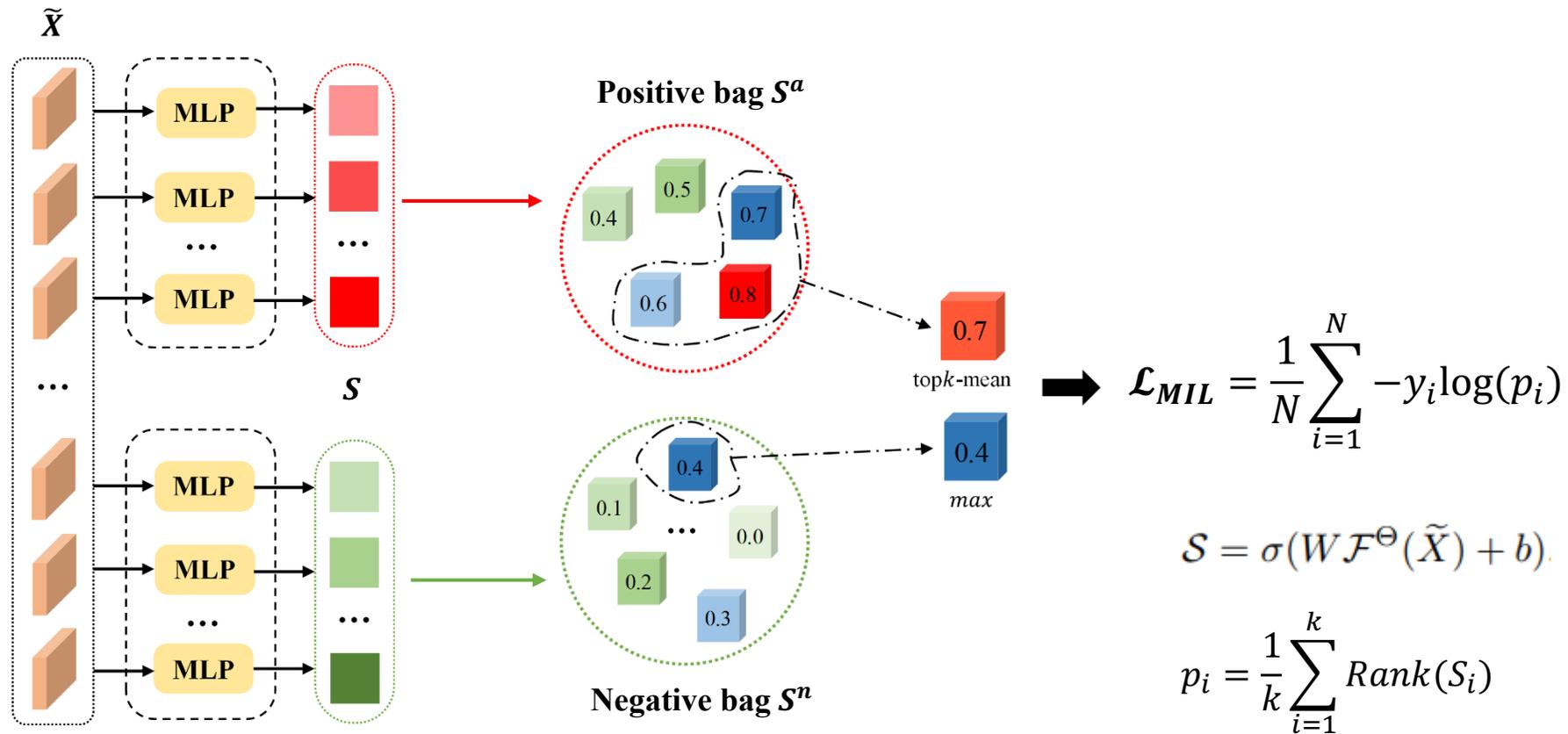
$$G_{ij} = \exp\left(-\frac{|i-j|^2}{2\sigma}\right)$$

$$\tilde{A} = A + G$$

$$\tilde{X} = \text{Norm}(\text{stack}(\tilde{A}XW_k)W_h + X)$$

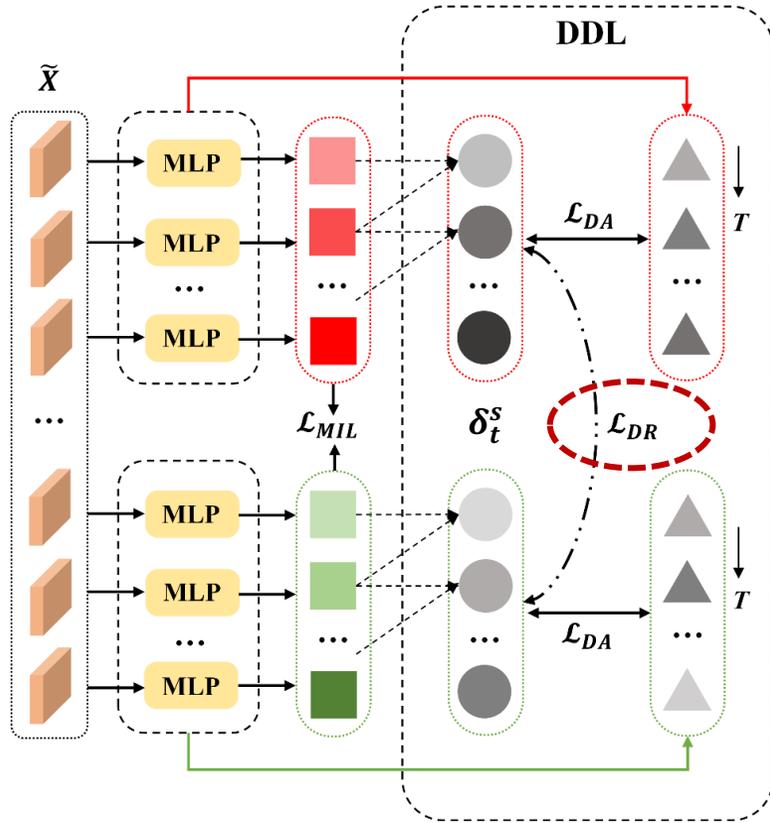
Methodology

Multiple Instance Learning



Methodology

Discriminative Dynamics Learning

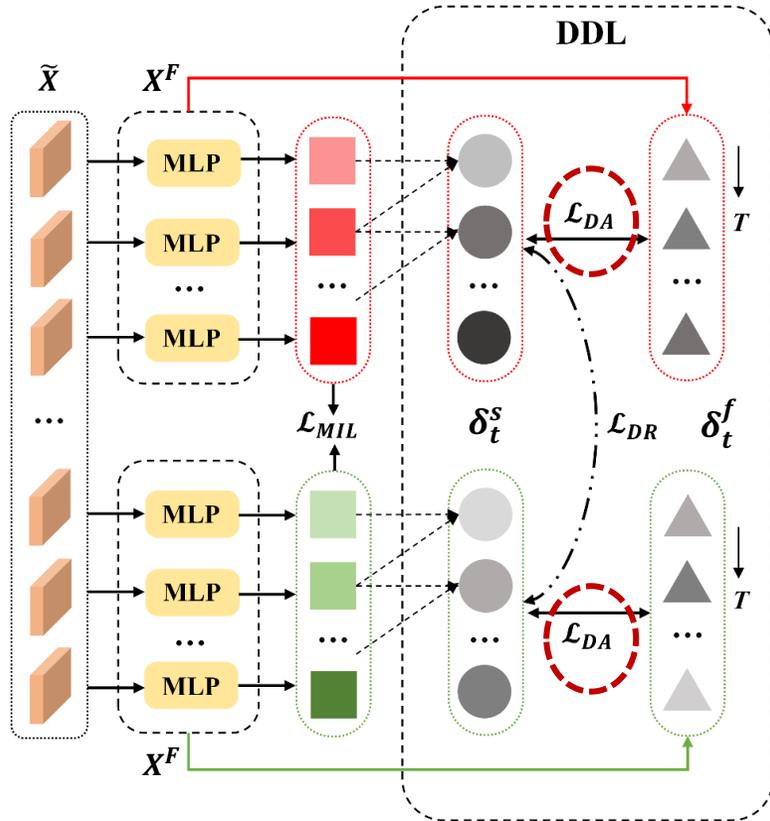


Score Dynamics Ranking --> Outer Bag

$$\begin{aligned}
 S &= \{s_1, s_2, \dots, s_t\} \\
 \delta_t^s &= |s_t - s_{t+1}| \\
 \varepsilon_{\Delta S} &= \frac{1}{k} \sum_{t=1}^k |\delta_t^s|^2
 \end{aligned}
 \left. \vphantom{\begin{aligned} S \\ \delta_t^s \\ \varepsilon_{\Delta S} \end{aligned}} \right\} \mathcal{L}_{DR} = \max(0, \zeta - \varepsilon_{\Delta S}^a + \varepsilon_{\Delta S}^n)$$

Methodology

Discriminative Dynamics Learning



Feature Dynamics Alignment --> Inner Bag

$$X^F = \{x_1^F, x_2^F, \dots, x_t^F\}$$

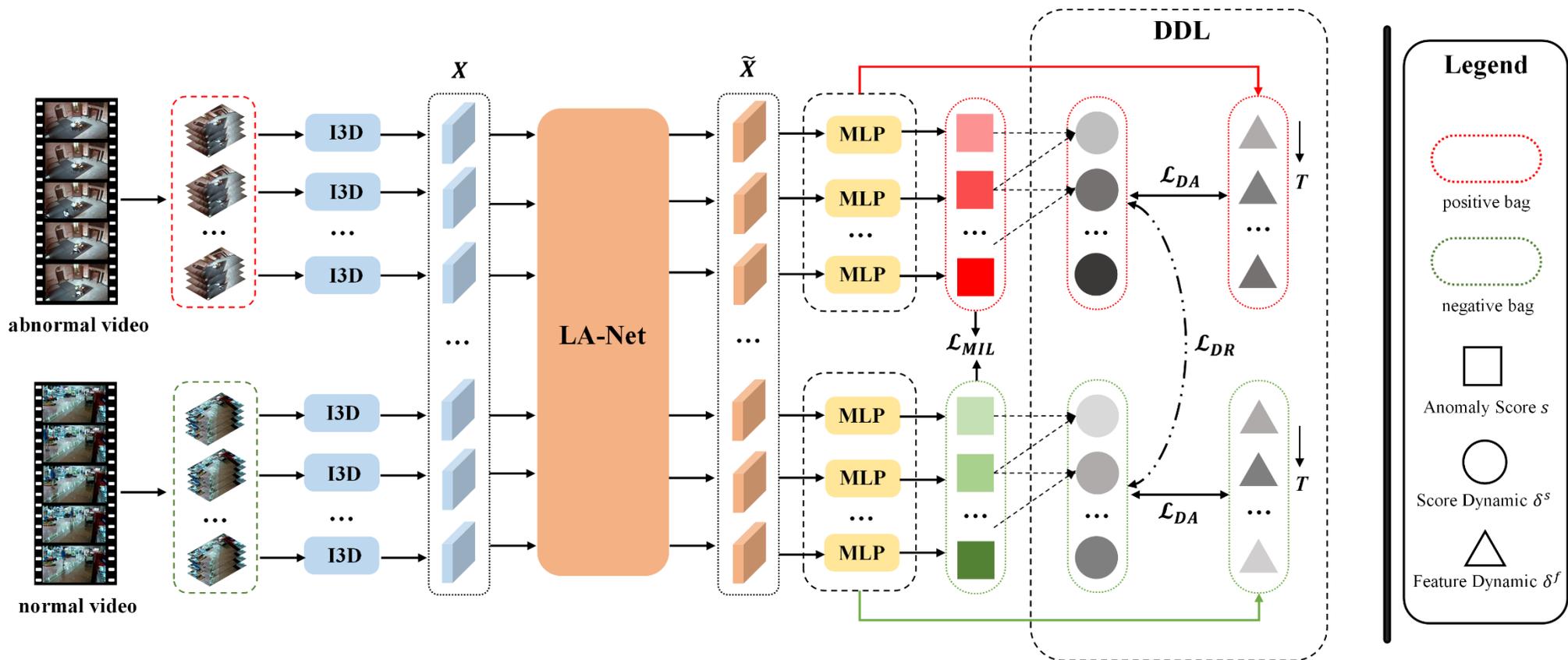
$$\delta_t^f = 1 - \frac{x_t^F x_{t+1}^F}{\|x_t^F\| \|x_{t+1}^F\|}$$

$$\delta_t^s = |s_t - s_{t+1}|$$

$$\mathcal{L}_{DA} = \frac{1}{N \times (T-1)} \sum_{i=1}^N \left(\sum_{t=1}^{T-1} -\delta_t^s \log(\delta_t^f + \epsilon) \right)_i$$

Methodology

Overall Objective Function



$$\mathcal{L} = \mathcal{L}_{MIL} + \lambda_1 \mathcal{L}_{DR} + \lambda_2 \mathcal{L}_{DA}$$

Experimental Results

State-Of-The-Art Performance

Table 1. Frame-level AUC performance on UCF-Crime.

| Method | Feature | AUC(%) |
|---------------------------|----------|--------------|
| Sultani <i>et al.</i> [9] | C3D RGB | 75.41 |
| Zhang <i>et al.</i> [10] | C3D RGB | 78.66 |
| Motion-Aware [21] | PWC Flow | 79.00 |
| Zhong <i>et al.</i> [11] | TSN RGB | 82.12 |
| Wu <i>et al.</i> [13] | I3D RGB | 82.44 |
| MS-BSAD [18] | I3D RGB | 83.53 |
| RTFM [20] | I3D RGB | 84.30 |
| DDL (Ours) | I3D RGB | 85.12 |

Table 2. Frame-level AP performance on XD-Violence.

| Method | Feature | AP(%) |
|---------------------------|---------|--------------|
| SVM baseline | - | 50.78 |
| OCSVM [22] | - | 27.25 |
| Hasan <i>et al.</i> [23] | - | 30.77 |
| Sultani <i>et al.</i> [9] | C3D RGB | 73.20 |
| Wu <i>et al.</i> [13] | I3D RGB | 75.41 |
| RTFM [20] | I3D RGB | 77.81 |
| DDL (Ours) | I3D RGB | 80.72 |

Experimental Results

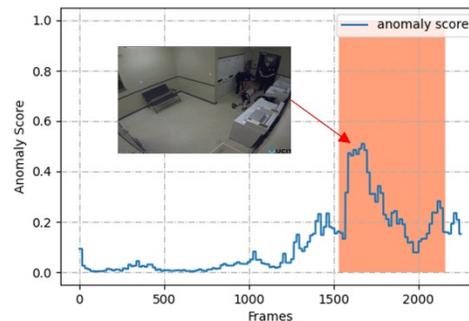
Ablation Study

Table 3. Ablation study of location prior.

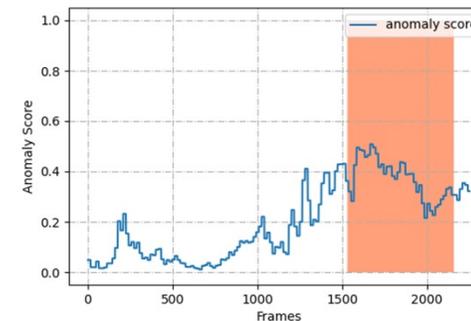
| Model | UCF-Crime AUC(%) | XD-Violence AP(%) |
|--------------------------------|---------------------|----------------------|
| LA-Net w/o prior \mathcal{G} | 83.06 | 78.41 |
| LA-Net w/ prior \mathcal{G} | 83.67 | 79.18 |

Table 4. Ablation study of the DDL method.

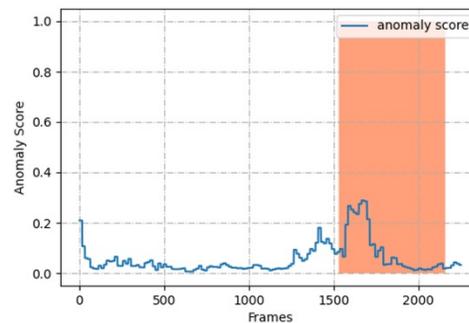
| \mathcal{L}_{MIL} | \mathcal{L}_{DR} | \mathcal{L}_{DA} | UCF-Crime AUC(%) | XD-Violence AP(%) |
|---------------------|--------------------|--------------------|---------------------|----------------------|
| ✓ | | | 83.67 | 79.18 |
| ✓ | ✓ | | 84.04 | 80.15 |
| ✓ | | ✓ | 84.33 | 80.23 |
| ✓ | ✓ | ✓ | 85.12 | 80.72 |



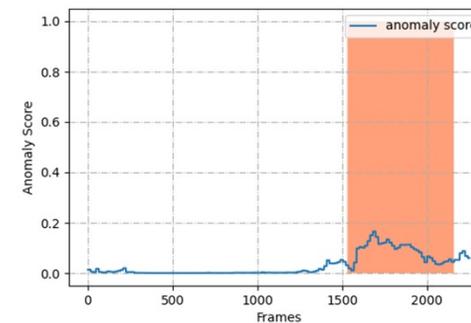
(a) \mathcal{L}_{MIL}



(b) $\mathcal{L}_{MIL} + \mathcal{L}_{DR}$



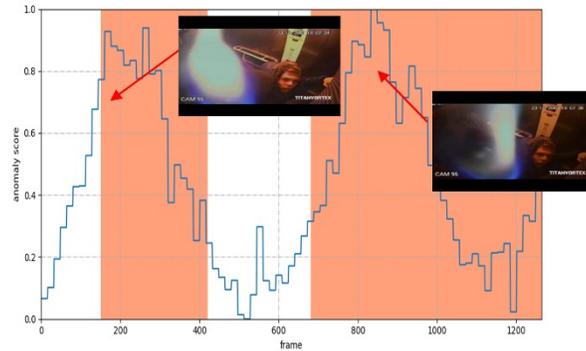
(c) $\mathcal{L}_{MIL} + \mathcal{L}_{DA}$



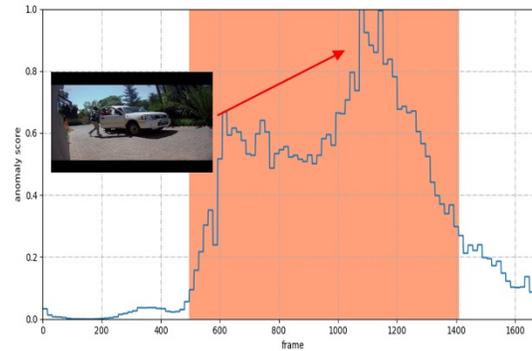
(d) $\mathcal{L}_{MIL} + \mathcal{L}_{DR} + \mathcal{L}_{DA}$

Qualitative Analysis

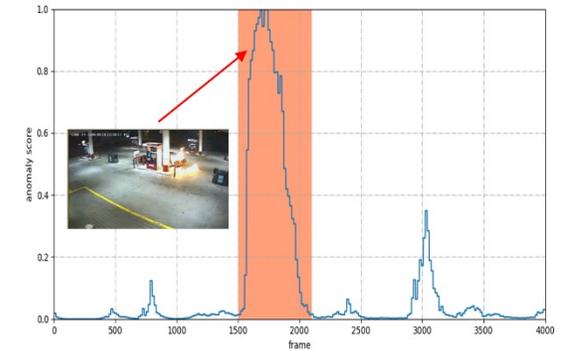
UCF-Crime



(a) Arson011

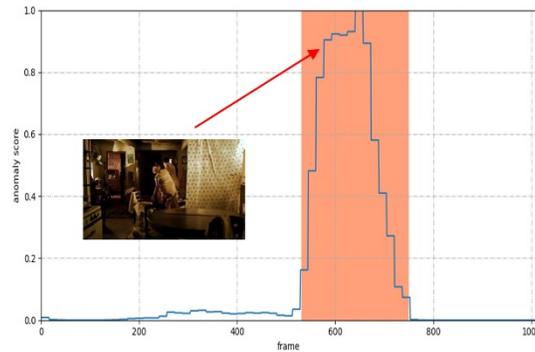


(b) Robbery050

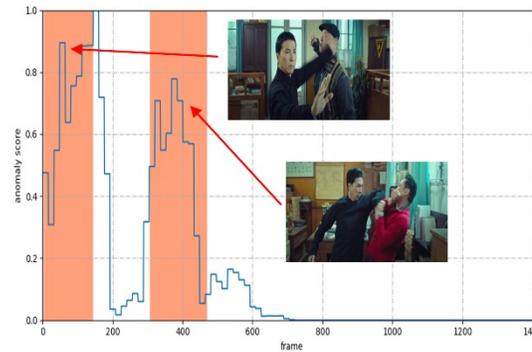


(c) Explosion002

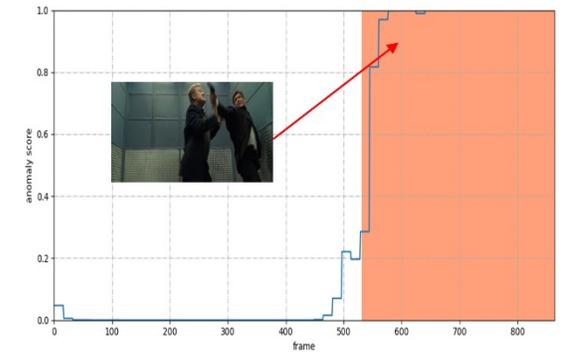
XD-Violence



(d) City.of.God.2002



(e) Ip.Man.3.2015



(f) Salt.2010

Thank You!