



Leveraging VLP models for cross-modal video retrieval

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Code is available:
https://github.com/yuqi657/ts2_net

Introduction

- Our method is based on recent image-language pre-trained model CLIP.
- We make adaptations to the origin visual transformer, to leverage VLP for video retrieval tasks.

Results

Main Results:

Team	Run #	xinfAP
C_D_RUCAIM3-Tencent.22	2	0.175
C_D_RUCAIM3-Tencent.22	1	0.119
C_D_RUCAIM3-Tencent.22	3	0.109
C_D_RUCAIM3-Tencent.22	4	0.094

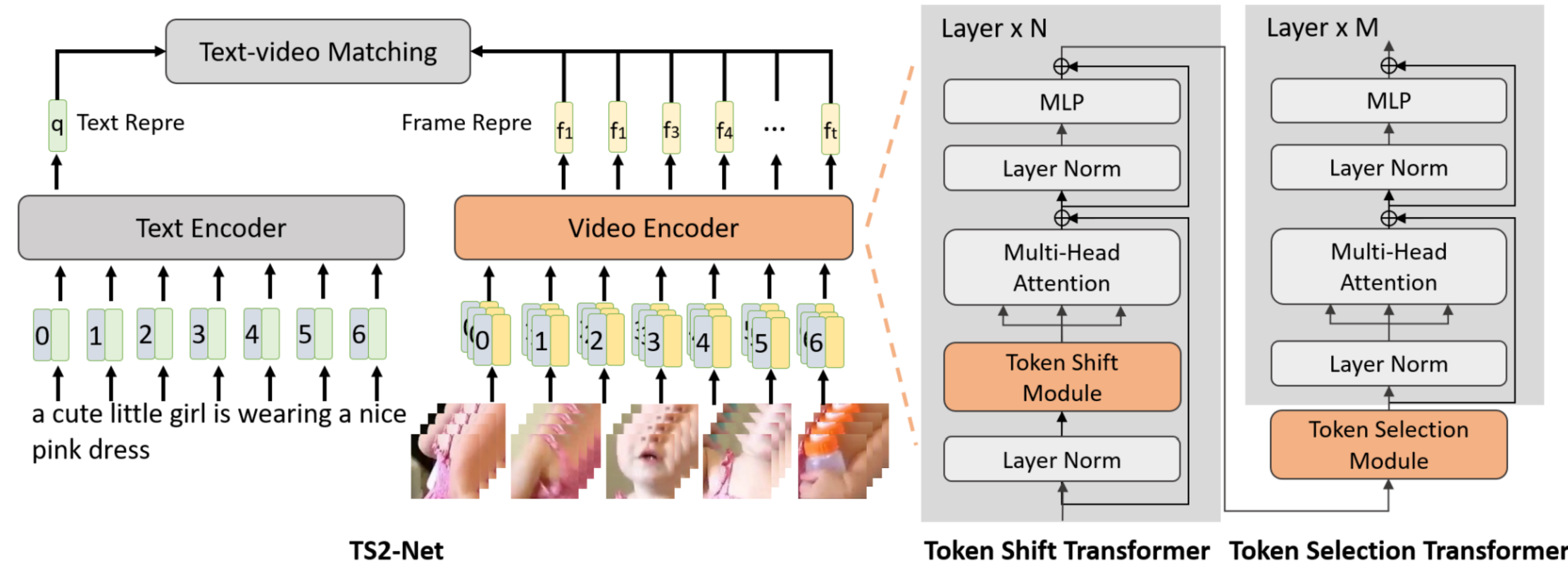
Novelty Score:

Team	Novelty	xinfAP
N_D_VIREO.22_6	38.4	0.088
C_D_RUCAIM3-Tencent.22_2	31.4	0.175
C_D_WasedaMeiseiSoftbank.22_2	26.7	0.282
C_D_kindai_ogu_osaka.22_1	23.1	0.199
C_D_RUCMM.22_2	18.4	0.262
C_D_ITI_CERTH.22_2	15.0	0.210
C_D_CamiloUchile.22_3	5.8	0.002

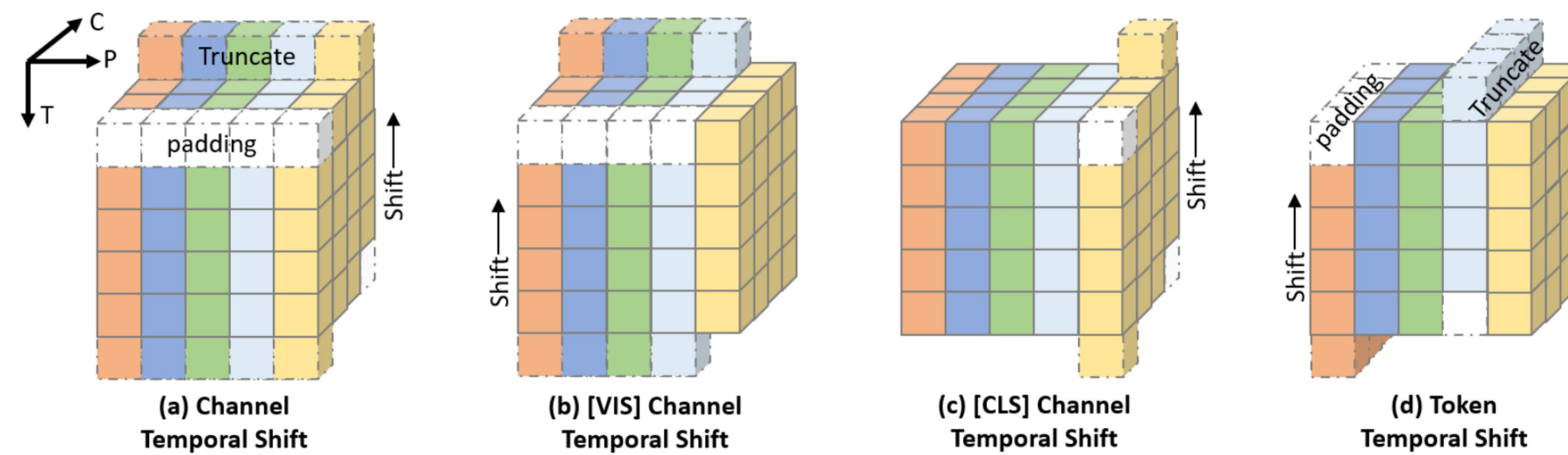
- Our best xinfAP is 0.175.
- Our method retrieves more unique relevant shots.

Method

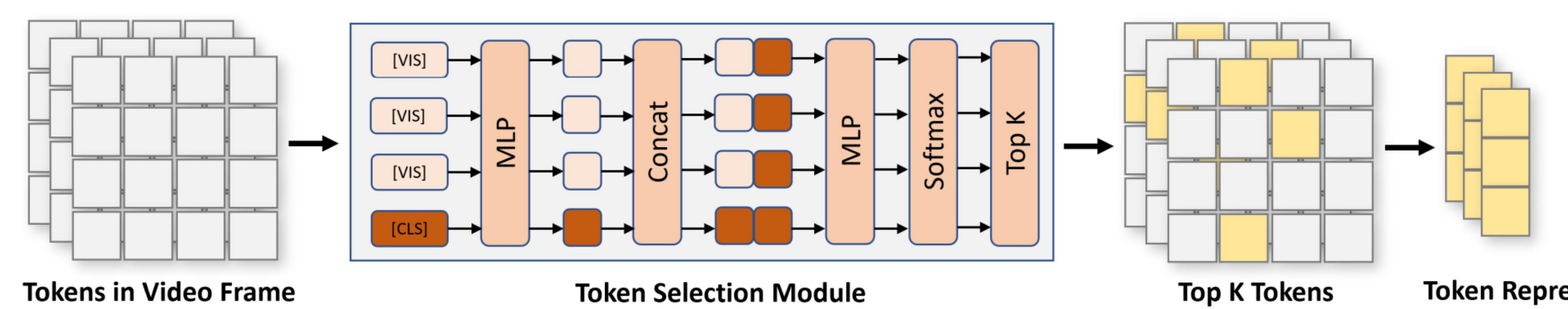
Main Architecture:



Token Shift Module:



Token Selection Module:



Matching, Training and Inference:

$$\text{Frame level sim: } s_i = \frac{q \cdot f_i}{\|q\| \|f_i\|}, \text{ Video level sim: } s = \sum_{i=1}^n \alpha_i s_i$$

Training Loss:

$$\mathcal{L}_t^{t2v} = -\frac{1}{B} \sum_i \log \frac{\exp(\tau \cdot \text{sim}(q_i, v_i))}{\sum_{j=1}^B \exp(\tau \cdot \text{sim}(q_i, v_j))}$$

$$\mathcal{L}_t^{v2t} = -\frac{1}{B} \sum_i \log \frac{\exp(\tau \cdot \text{sim}(q_i, v_i))}{\sum_{j=1}^B \exp(\tau \cdot \text{sim}(q_j, v_i))}$$

$$\mathcal{L} = \frac{1}{2} (\mathcal{L}_{t2v} + \mathcal{L}_{v2t})$$

Analysis

Queries with great results:

- 704 A parked white car
- 726 Two teams playing a game where one team have their players wearing white t-shirts



Queries with bad results:

- 702 A room with blue wall
- 713 A kneeling man outdoors
- 728 Two adults are seated in a flying paraglider in the air



- Our method retrieves the correct elements in each query (e.g. blue, wall, paraglider, etc), but fails to model relations between these elements.
- Some elements are difficult to identify (e.g. kneeling). This might be caused by the domain of pre-trained data.

Puzzle

Q1: Are there some annotation errors or misunderstanding of some concepts?

- 711 A woman wearing a head kerchief (0 in top10)
- 730 A man is holding a knife in a non-kitchen location (3/10)



Q2: Is there possible to design a metric considering both xinfAP and diversity?