

Temporal CFT: Multi-Temporal Cross-Modality Fusion Transformer for Multispectral Video Object Detection

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Motivation

Challenges:

- 1. Fusion** - RGB sensors can suffer in poor lighting and weather conditions, whereas thermal cameras operate well in poor lighting and weather conditions but can have a lower resolution.
- 2. Temporal** - Single-frame models fail to consider temporal relationship in traffic object movement.

Goal: Multi-temporal Cross-Modality Fusion for Thermal + RGB sensors

Problem Statement

Input:
 RGB video sequence $I_R \in \mathbb{R}^{HW \times C \times N}$
 + Thermal video sequence $I_T \in \mathbb{R}^{HW \times C \times N}$

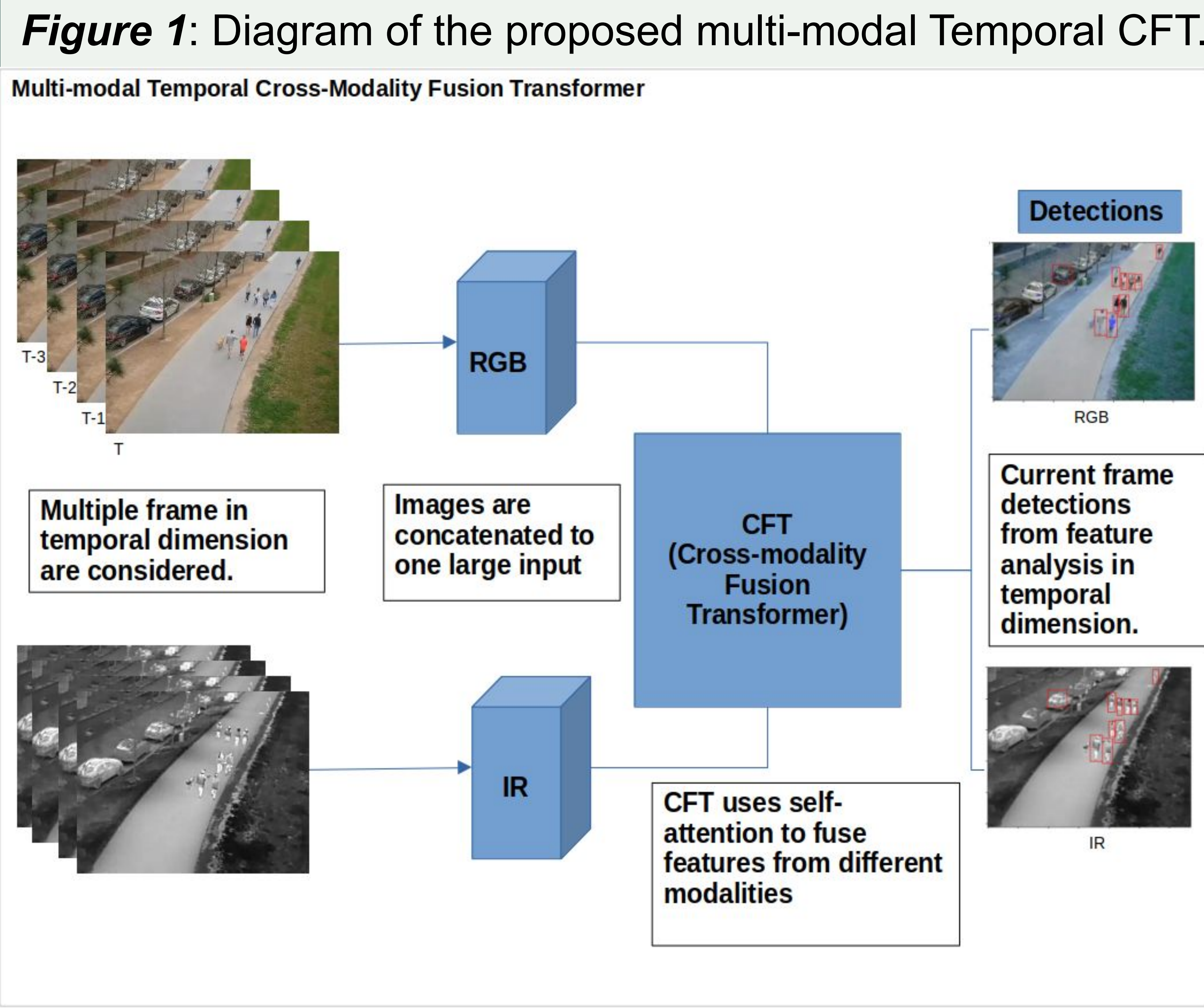
Output:
 Object bounding boxes at last frame

Network:
 Cross-Modality Transformer [3] for fusion with temporal concatenation; YOLOv5 detector

References

[1] A. Berg, J. Ahlberg, and M. Felsberg, "A thermal object tracking benchmark," in Int. Conf. Advanced Video and Signal Based Surveillance, 2015, pp. 1–6.
 [2] V. Vs, J. M. J. Valanarasu, P. Oza, and V. M. Patel, "Image fusion transformer," in IEEE Int. Conf. Image Processing, 2022, pp. 3566–3570.
 [3] F. Qingyun, H. Dapeng, and W. Zhaokui, "Cross-modality fusion transformer for multispectral object detection," arXiv preprint arXiv:2111.00273, 2021.
 [4] A. Vaswani, N. Shazeer, N. Parmar, J. Uszkoreit, L. Jones, A. N. Gomez, Ł. Kaiser, and I. Polosukhin, "Attention is all you need," NIPS, vol. 30, 2017.
 [5] E. Gebhardt and M. Wolf, "Camel dataset for visual and thermal infrared multiple object detection and tracking," in Int. Conf. Advanced Video and Signal-based Surveillance, 2018.

Method



Takeaways

1. Temporal model achieved 78% True positive rate compared to 54% from baseline: **24% improvement**
2. Temporal models showed **better performance in handling data imbalance** than baseline.
3. **In future:** Our research can be focused on designing and implementing fusion in the temporal dimension that is more memory efficient and better utilizes multi-modal features.
4. Temporal models trained with multi-modalities are better in **learning, inference and classification.**

Results

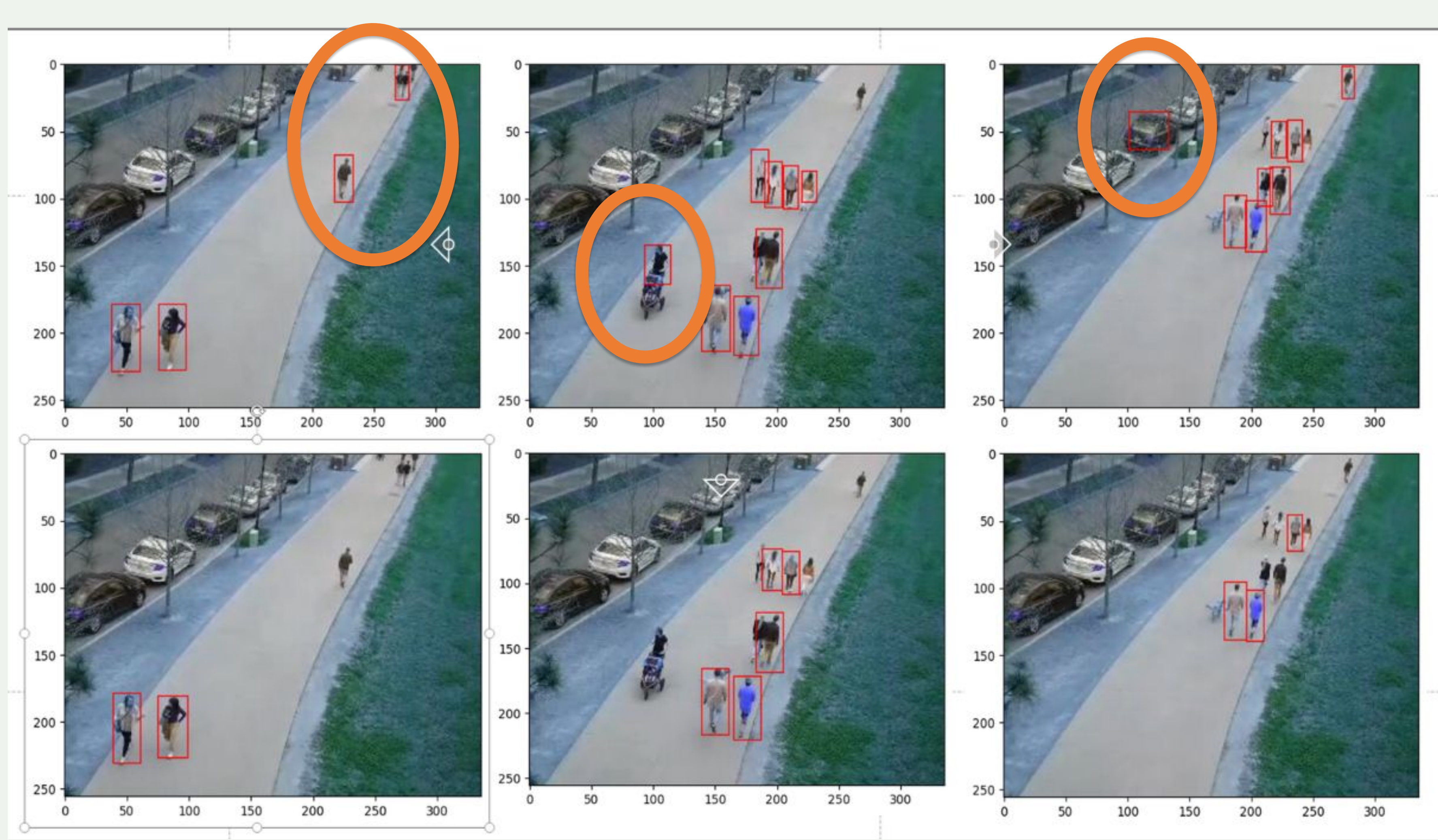


Figure 2: Object Detection Results. Top: Temporal-CFT model (proposed), Bottom: Single-frame model (Baseline). **Orange** indicates correct detections from the proposed Temporal CFT.

Table 1: Person detection results on the CAMEL dataset.

Model Type	True Positive ↑	False Negative ↓
Single-Frame CFT	54	46
Temporal CFT (Proposed)	78	22

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