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Brisbane, Australia 6-10 April

A Cascade Transformer-Based Multi-Scale Framework for Object Detection and Instance Segmentation in Remote Sensing Imagery

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BACKGROUND AND INTRODUCTION



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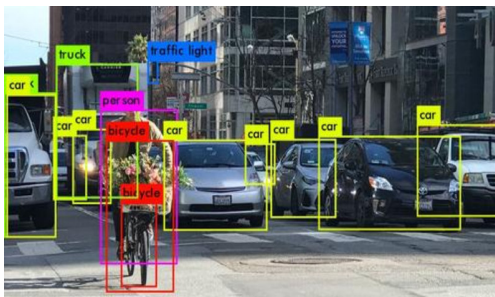
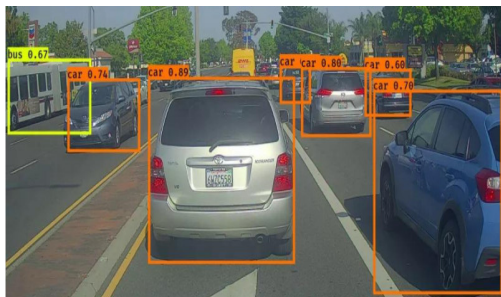


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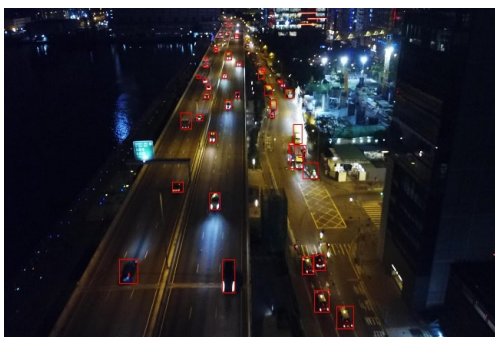


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Object detection in natural images



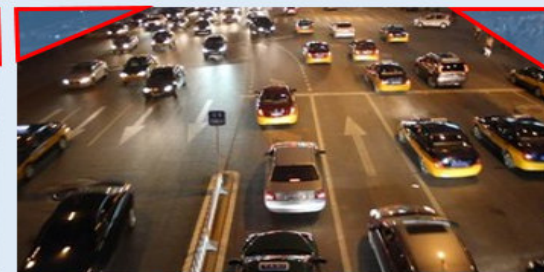
Object detection in RS images



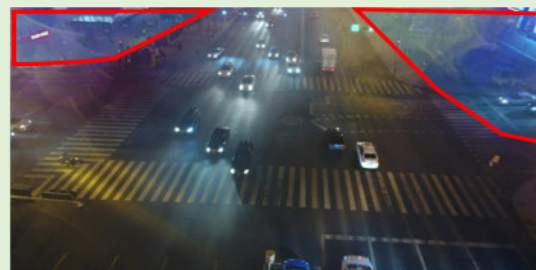
Change detection in RS images



Background regions in Natural Images



Background regions in Remote Sensing Images



Research background

Current deep learning-based change detection methods can be categorized into **pixel-level** and **object-level** methods

Pixel-level methods



Can obtain **high detection accuracy**, but it is difficult to **distinguish each change object** when objects are densely distributed

Object-level methods



Can **distinguish changed objects**, but it is difficult to **obtain accurate boundary representation**

Motivation

Developing fine-grained object-level change detection with accurate boundary and distinguishing individual instances

Research background

Current deep learning-based change detection methods can be categorized into **pixel-level** and **object-level** methods

Therefore, we propose a Cascade Transformer-based Multi-Scale Framework

Motivation

Developing fine-grained object-level change detection with accurate boundary and distinguishing individual instances



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THE PROPOSED METHOD



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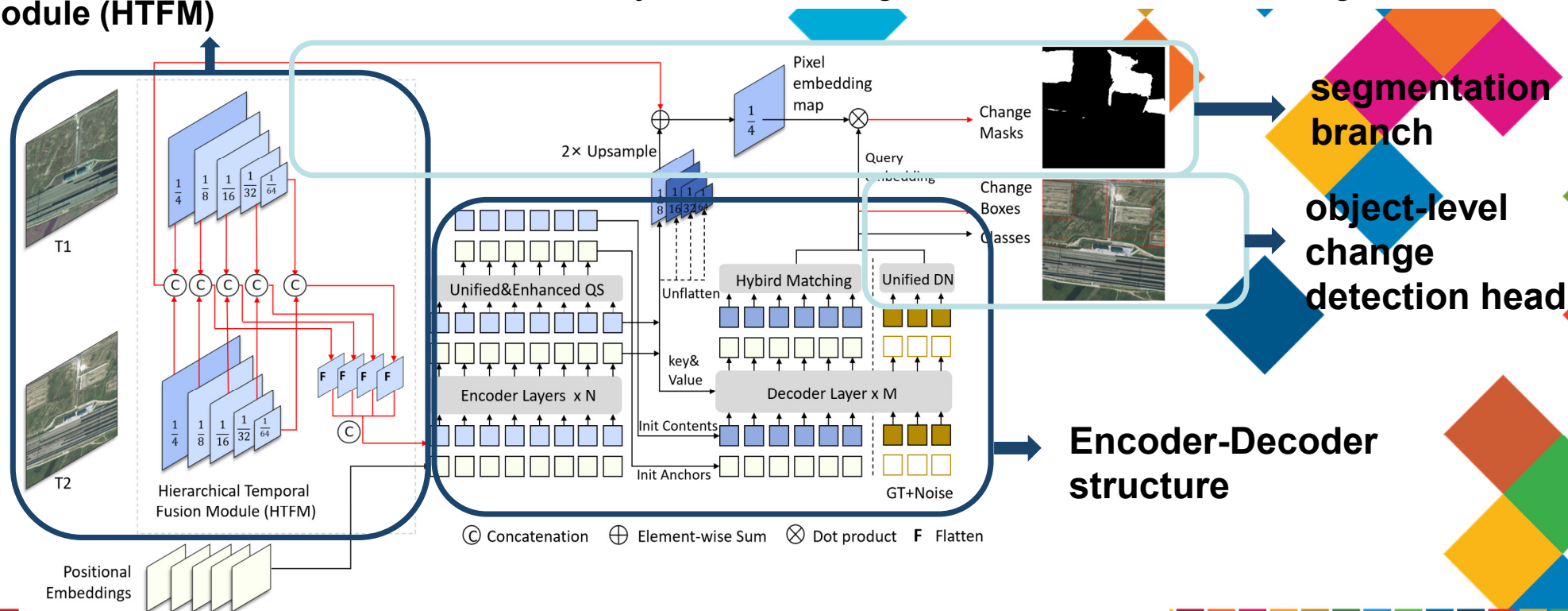
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ARCHITECTURE

Hierarchical Temporal Fusion Module (HTFM)

This framework comprises a Hierarchical Temporal Fusion Module (HTFM), Transformer-based Encoder-Decoder, an object-level change detection head, and a segmentation branch



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HTFM: Used to extract and fuse multi-scale features, the formula is expressed as

$$F_{bi} = \text{Concatenate}(F_{bi}^1, F_{bi}^2), (i = 1, 2, 3, 4, 5)$$

$$F_{hi} = \text{Flatten}(F_{bi}), (i = 2, 3, 4, 5)$$

$$F_e = \text{Concatenate}(F_{hi}), (i = 2, 3, 4, 5)$$

Loss Function

Includes a localization loss and a classification loss for object-level change detection, as well as a Mask loss for segmentation tasks

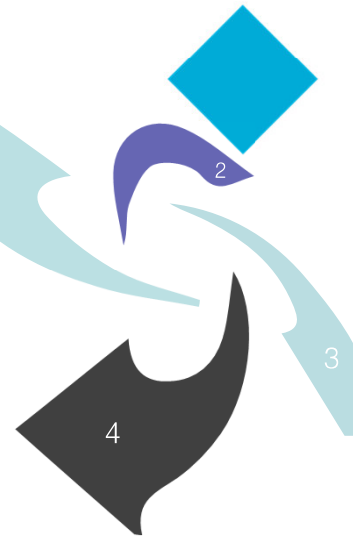
$$L_{hibird} = \lambda_{cls} L_{cls} + \lambda_{L1} L_{L1} + \lambda_{GIOW} L_{GIOW} + \lambda_{ce} L_{ce} + \lambda_{dice} L_{dice}$$

Encoder-Decoder structure based on Transformer

Used to get predictions for box and mask initialization contents and anchor box queries

Object-level change detection head and Segmentation branch

Obtain box representations of changed regions and fine-grained boundary representations





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The proposed Cascade Transformer-based Multi-Scale Framework

01

The **FIRST** transformer-based object-level D&CD framework

Problems: Transformer-based CD methods are hard to train; existing methods lack precision.

Our Method: Inspired by the succeed models in CV field, effectively achieving transformer-based object-level change detection.

02

the **FIRST** unified object-level change detection and segmentation framework

Problems: Current methods output bboxes only, which are imprecise.

Our Method: Outputs results with the bbox and the fine boundary masks, and achieves better performance even better than pixel-level methods.

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EXPERIMENTS AND RESULTS



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Dataset <https://github.com/xiaoxiangAQ/LIM-CD-dataset>

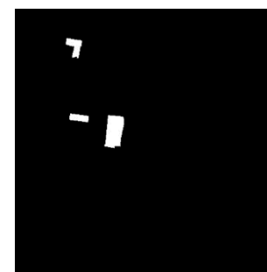
LIM-CD: a large-scale high-resolution 2D change detection benchmark dataset, consists of 9,259 pairs of pre- and post-temporal high resolution images, divided into a training set with 6,547 pairs, a validation set with 1,776 pairs, and a test set with 936 pairs.



pre-temporal images



post-temporal images



label

Image sizes: 512x512 pixels

ground sampling distance ranging: from 0.5 to 2 meters

We compared our experimental results with the following SOTA methods:

(1) Transformed based pixel-level change detection methods include BIT-CD and ChangeFormer.

(2) Other CNN-based pixel-level change detection methods include FCEF, FC-Siam-diff, FC-Siam-conc, ISNet, SUNET_EP50 and SUNET.

The dual output mode (box and mask) of the proposed framework addresses the challenge of comparing object-level and pixel-level change detection methods.



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Experimental results of different methods on the LIM-CD dataset*

Method	Precision	Recall	IOU	F1
CNN-based pixel-level change detection methods				
FCEF	64.87	54.47	42.06	59.22
FC-Siam-diff	66.29	52.41	41.38	58.54
FC-Siam-conc	64.54	46.92	37.30	54.34
ISNet	66.41	54.63	42.80	59.95
SNUNET_EP50	72.01	55.99	45.98	63.00
SNUNET	73.27	57.19	47.31	64.24
Transformer-based pixel-level change detection methods				
BIT-CD	74.34	51.05	43.40	60.53
ChangeFormer	70.84	45.36	38.22	55.31
Our Method	67.30	64.01	48.83	65.62

*All values in the table are expressed as percentages (%)

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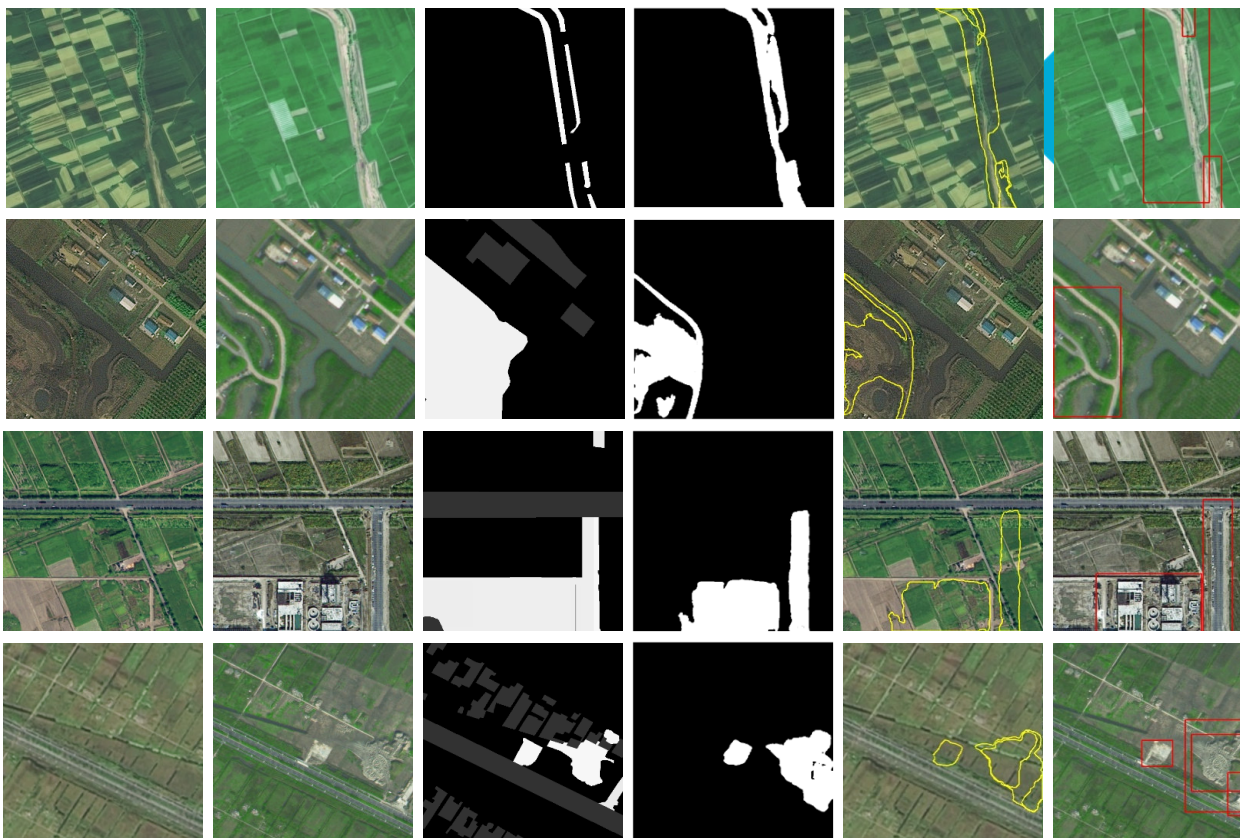
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Partial visualization of the proposed framework



(a) T1 image

(b) T2 image

(c) Ground Truth(GT)

(d) binary results from the
change area Mask

(e) change area Mask
(present on the former phase)

(f) change area box
(present on the later phase)



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Received third place in the competition supported by the National Natural Science Foundation of China, based on the enhanced proposed method with several additional strategies:

- **Two-stage Progressive Training:** Solves knowledge transfer in diverse scenes; 3 hours total training.
- **Rich Data Augmentation Techniques:** Significantly improve model generalization in challenging scenarios.
- **Efficient Inference:** Multi-process and multi-batch design boosts efficiency.



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ISPRS International Individual Tree Crown (ITC) Segmentation Contest, which attracted **over 40 teams** and around **200 participants** from **13 countries and regions**, including China, the United States, Canada, and France.

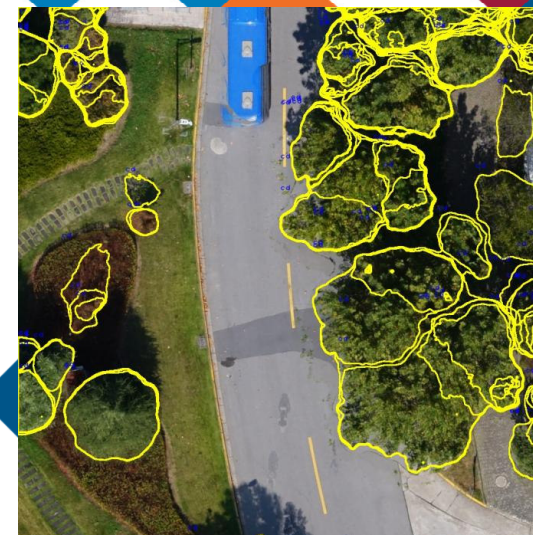
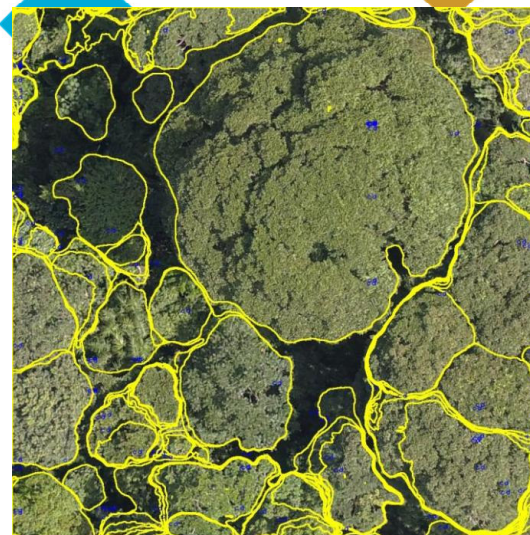
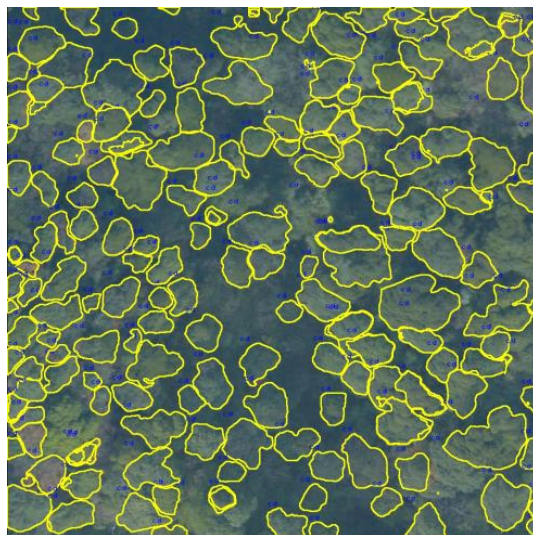
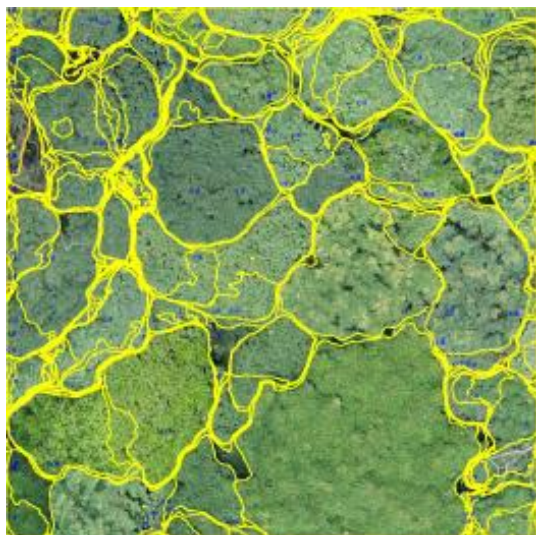
The competition ran from **January 29 to June 22, 2024**, and included two stages:
a **ranking stage** and a **final evaluation stage**.

- **Changed the dual-branch temporal input into a single-branch structure, using one-time remote sensing imagery.**
- **Introduced a lightweight Feature Pyramid Network (FPN) to better align multi-scale features across the network.**



Received Golden Prize (1st) in the ISPRS International Contest on Individual Tree Crown (ITC) Segmentation, based on the proposed framework with several additional strategies.

Experiment Results





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CONCLUSIONS

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A Cascade Transformer-Based Multi-Scale Framework for Object Detection and Instance Segmentation in Remote Sensing Imagery

- We proposed a cascade Transformer-based multi-scale framework for object detection and instance segmentation in remote sensing imagery.
- The method integrates object-level detection and mask-level segmentation in a unified structure, and handles complex scenes with varying object scales.
- Originally designed for change detection, the framework was successfully adapted to single-image tasks, and achieved first place in the ISPRS ITC Segmentation Contest.



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