



Collaboration, Innovation and Resilience: Championing a Digital Generation

Brisbane, Australia 6-10 April

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

Authors: Ruiqian Zhang, Qin Yan, Hanchao Zhang, Xiaogang Ning

Affiliations: Institute of Photogrammetry and Remote Sensing, Chinese Academy of Surveying and Mapping, China



PLATINUM SPONSORS





WORKING  
WEEK 2025

AND

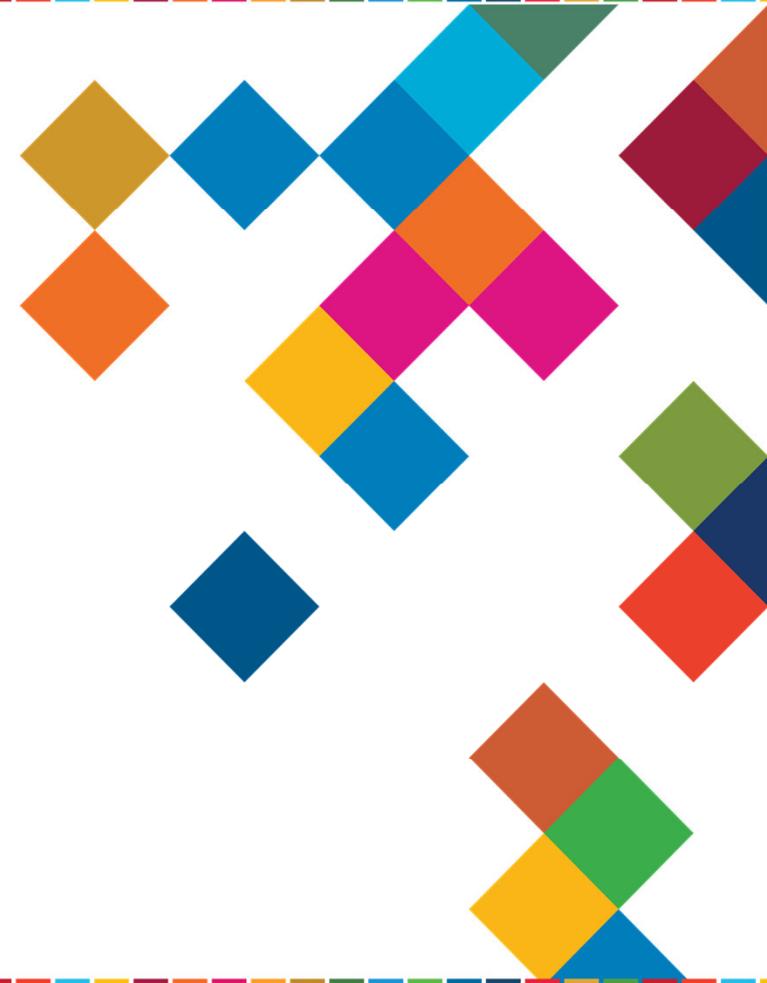
Locate25 | G  
THE NATIONAL GEOSPATIAL CONFERENCE

Collaboration, Innovation and Resilience:  
Championing a Digital Generation



Brisbane, Australia 6–10 April

1. Background and Introduction
2. The Proposed Method
3. Experiments and Results
4. Conclusions



ORGANISED BY



PLATINUM SPONSORS





**WORKING  
WEEK 2025**

AND

**Locate25** | **G**  
THE NATIONAL GEOSPATIAL CONFERENCE

Collaboration, Innovation and Resilience:  
Championing a Digital Generation



Brisbane, Australia 6–10 April

# BACKGROUND AND INTRODUCTION



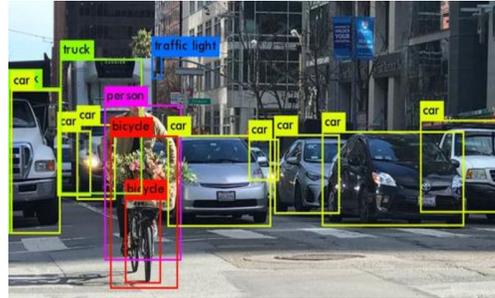
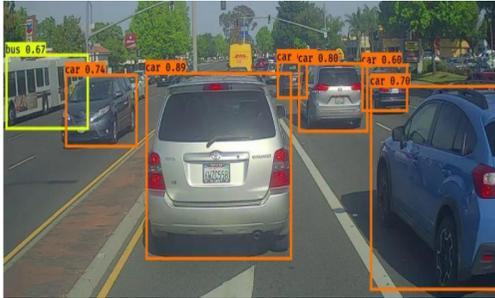
ORGANISED BY



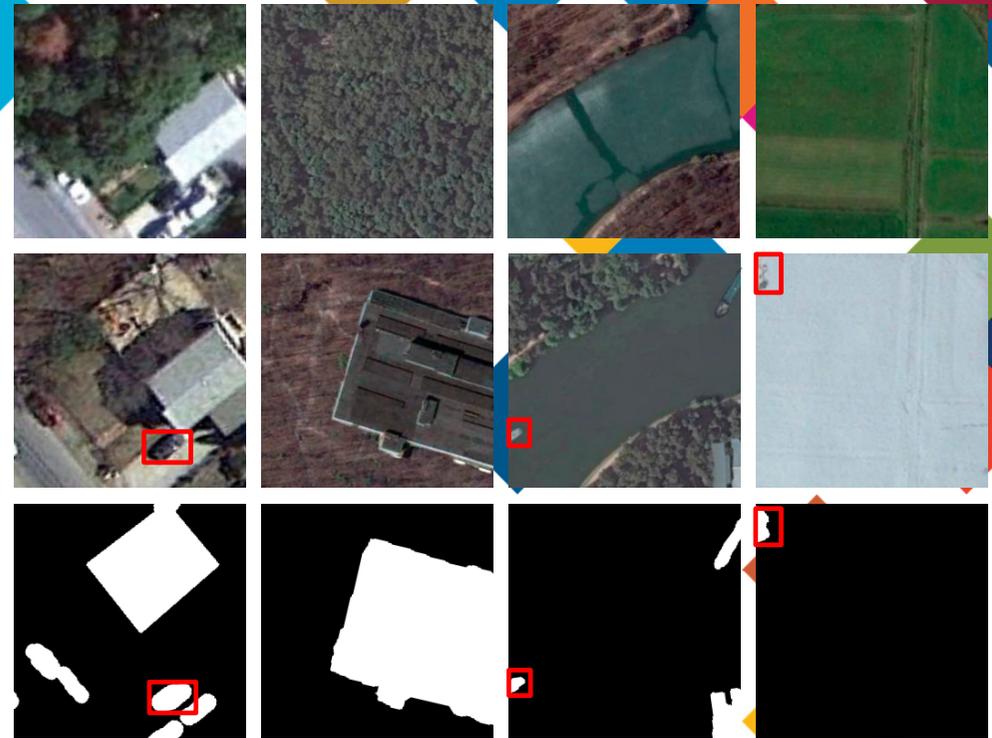
PLATINUM SPONSORS



## Object detection in natural images



## Change detection in RS images



## Object 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



**WORKING  
WEEK 2025**

AND

**Locate25** | **G**  
THE NATIONAL GEOSPATIAL CONFERENCE

Collaboration, Innovation and Resilience:  
Championing a Digital Generation



Brisbane, Australia 6–10 April

# THE PROPOSED METHOD



ORGANISED BY



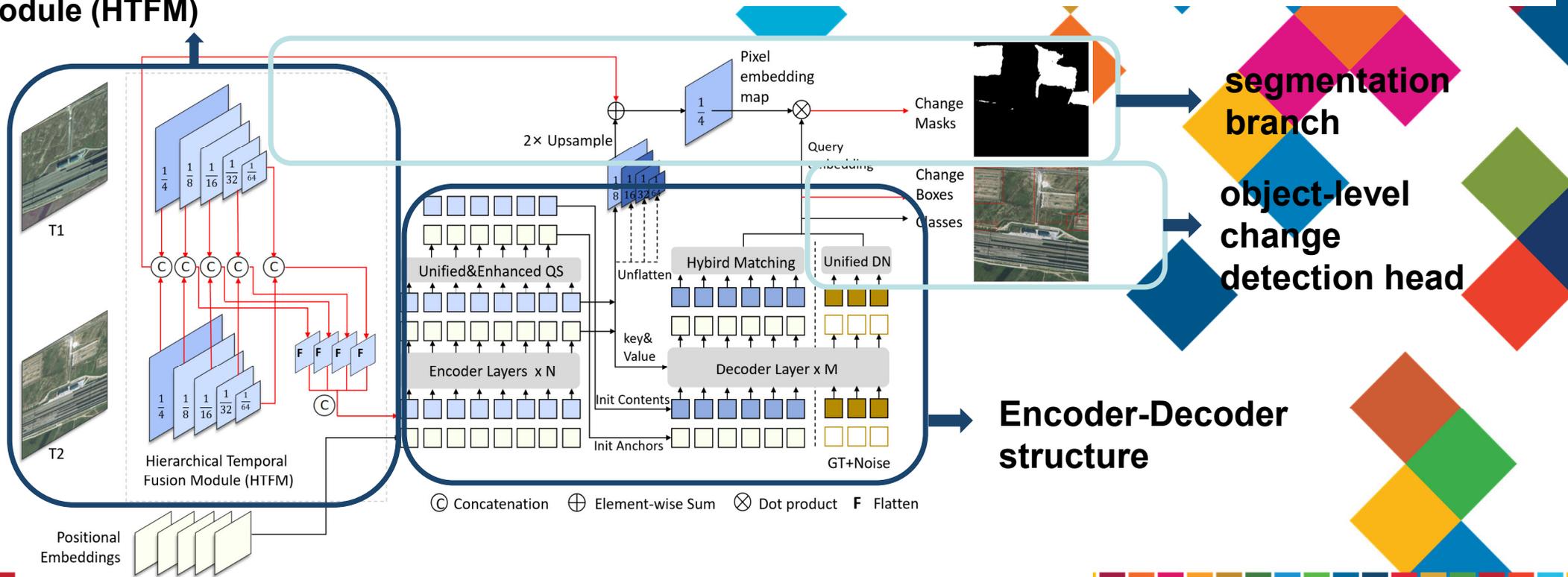
PLATINUM SPONSORS



# 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



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

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

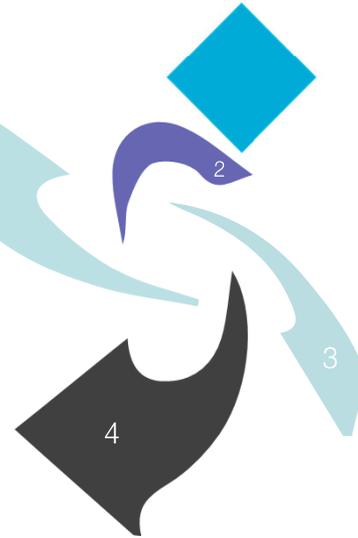
$$F_{hi} = \text{Flatten}(F_{bi}), (i = 2, 3, 4, 5)$$

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

### 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



# 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.

## 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.

02



**WORKING  
WEEK 2025**

AND

**Locate25**  
THE NATIONAL GEOSPATIAL CONFERENCE



Collaboration, Innovation and Resilience:  
Championing a Digital Generation



**Geospatial**  
Council of Australia

Brisbane, Australia 6–10 April

# EXPERIMENTS AND RESULTS



ORGANISED BY



**Geospatial**  
Council of Australia

PLATINUM SPONSORS



Australian Government



**esri**  
THE SCIENCE OF WHERE™



**HD  
Meter**



**Surveyors**  
Australia

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

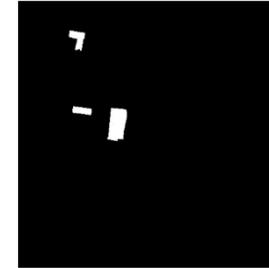
**LIM-CD**: a large-scale high-resolution 2D change detection benchmark dataset, consists of 9,259 pairs of pre- and post-temporal high resolution images, divided into a training set with 6,547 pairs, a validation set with 1,776 pairs, and a test set with 936 pairs.



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.**

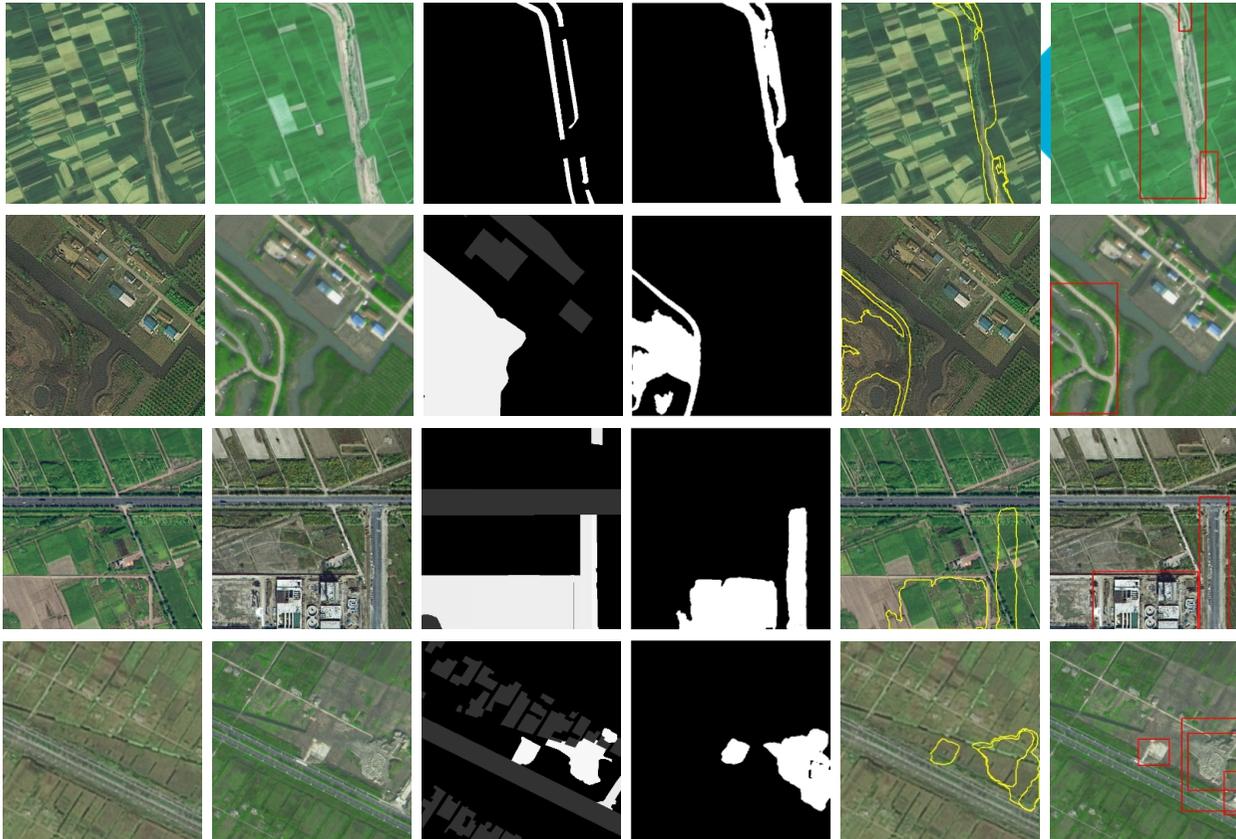


### Experimental results of different methods on the LIM-CD dataset\*

Method	Precision	Recall	IOU	F1
<b>CNN-based pixel-level change detection methods</b>				
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
<b>Transformer-based pixel-level change detection methods</b>				
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 (%)

## 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)

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.



**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.

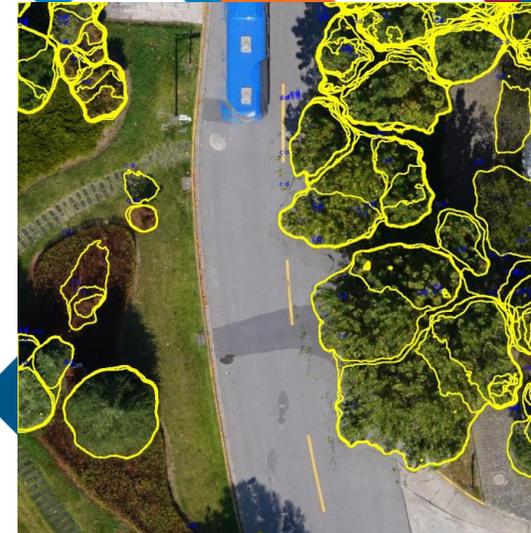
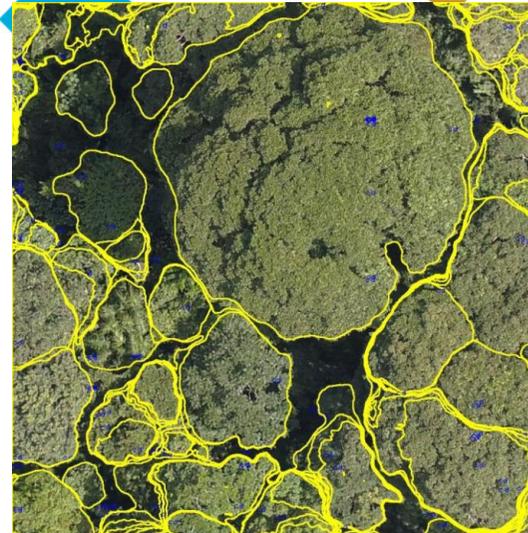
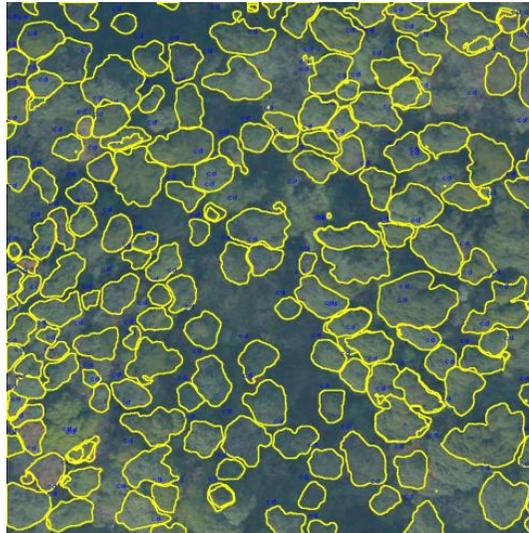
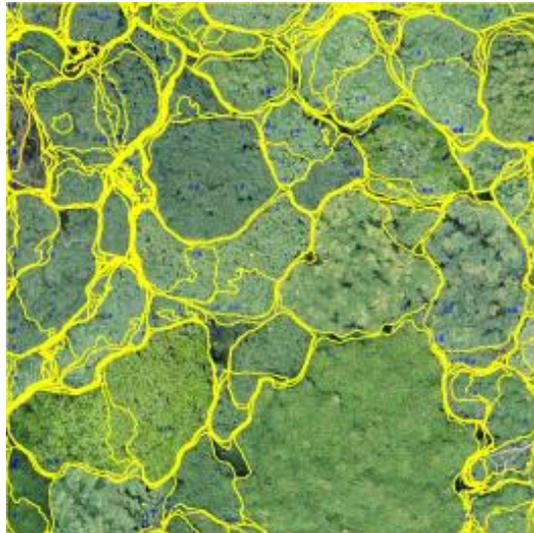
The competition ran from **January 29 to June 22, 2024**, and included two stages: a **ranking stage** and a **final evaluation stage**.

- **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 (1<sup>st</sup>) in the ISPRS International Contest on Individual Tree Crown (ITC) Segmentation, based on the proposed framework with several additional strategies.

# Experiment Results





**WORKING  
WEEK 2025**

AND

**Locate25** | **G**  
THE NATIONAL GEOSPATIAL CONFERENCE

Collaboration, Innovation and Resilience:  
Championing a Digital Generation



Brisbane, Australia 6–10 April

# CONCLUSIONS



ORGANISED BY



PLATINUM SPONSORS



# 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.



- **Name:** Ruiqian Zhang
- **Institution:** Chinese Academy of Surveying and Mapping
- **Academic Title:** Associate Research Professor
- **Degree:** PhD in Engineering
- **Research Interests:** Image processing, computer vision, remote sensing, deep learning
- **E-mail:** [zhangrq@casm.ac.cn](mailto:zhangrq@casm.ac.cn);  
[zhangruiqian@whu.edu.cn](mailto:zhangruiqian@whu.edu.cn)



Homepage