

# FSG-Net: Exploring Frequency-Spatial Synergy for Remote Sensing Change Detection



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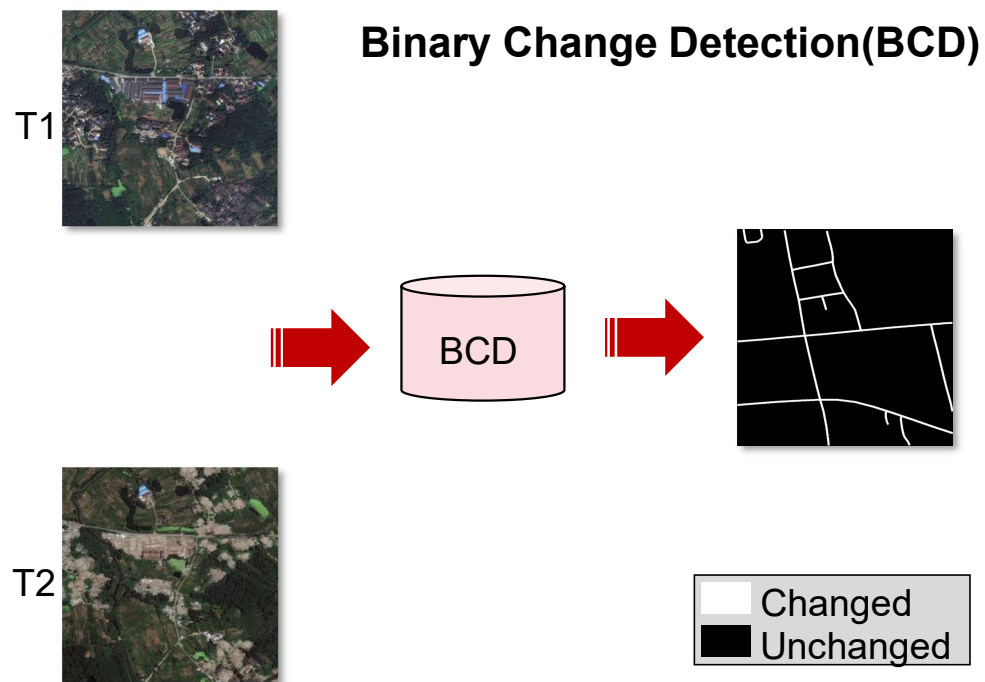
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# 1. Background: What is Change Detection?

Change Detection in high-resolution remote sensing images is a vital technology for tracking geospatial changes at different times.



Urban Expansion



Resource Management



Disaster Assessment



Environmental Monitoring

**A fundamental task for understanding the Earth's surface dynamics**

# 2. Motivation: Limitations of Existing Methods

■ True Negative      □ True Positive      ■ False Negative      ■ False Positive

**Challenge1:** Pseudo-changes caused by discrepancies in imaging conditions



+ (DAWIM&STSAM)

**Challenge2:** Imprecise boundary delineation resulting from the semantic gap between multilevel features



+ LGFU  
**Ours**

T1

T2

Labels

Others

# 3. Method: Overall Architecture

## Frequency-Spatial synergistic Gated network for CD task (FSG-Net)

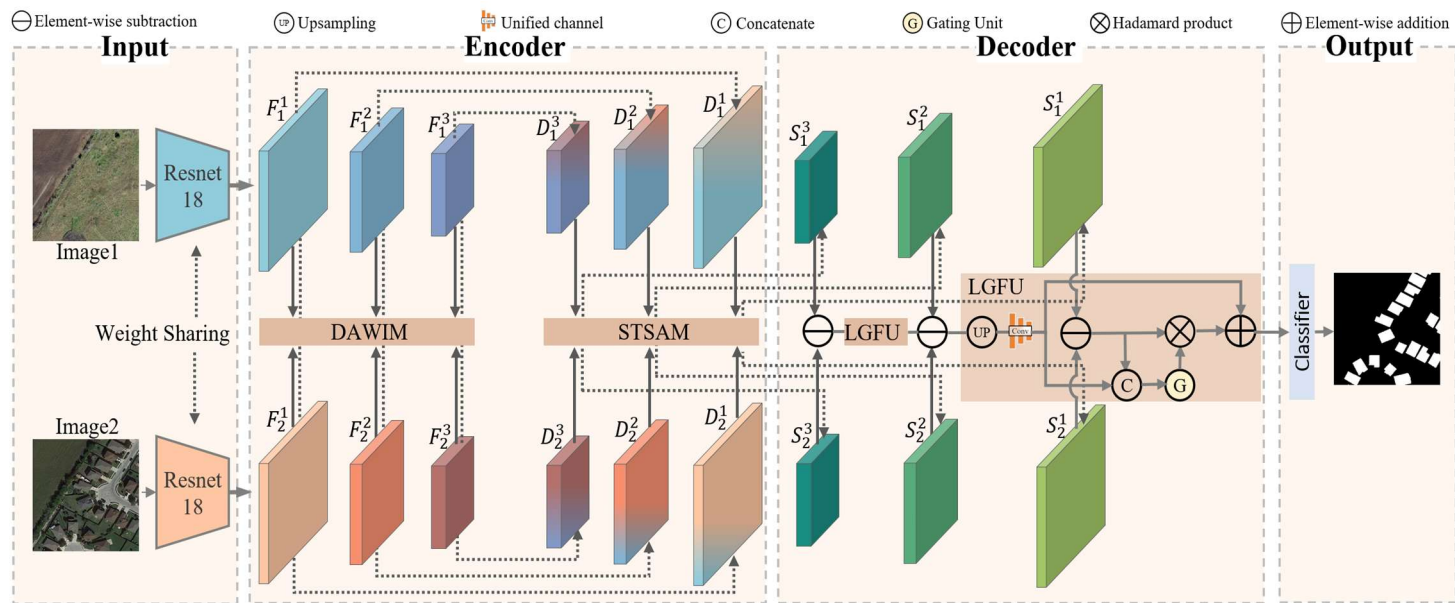
Frequency Interaction



S-T Attention



Gating Fusion



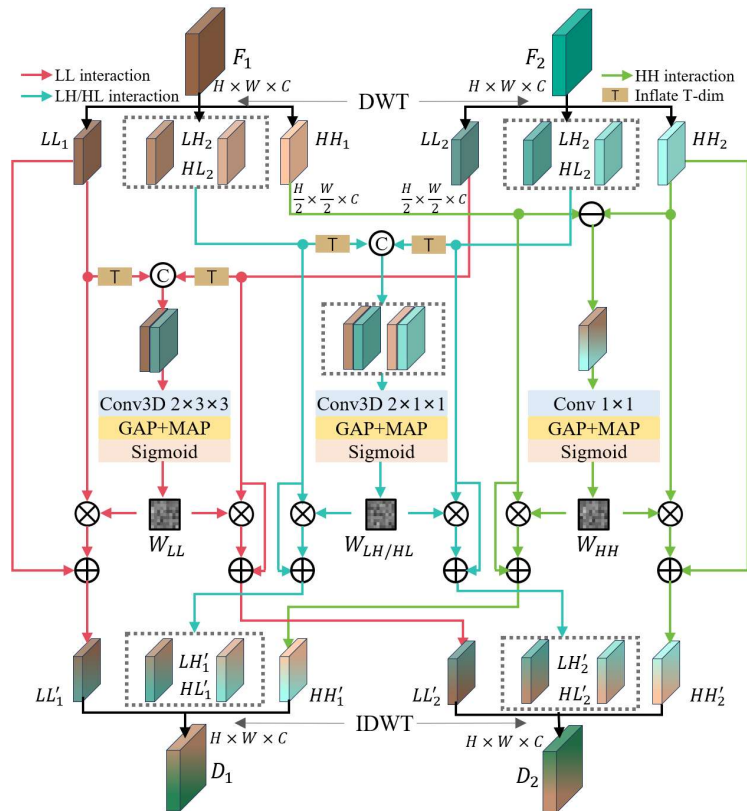
Overview of the proposed FSG-Net architecture

### Key Ideas:

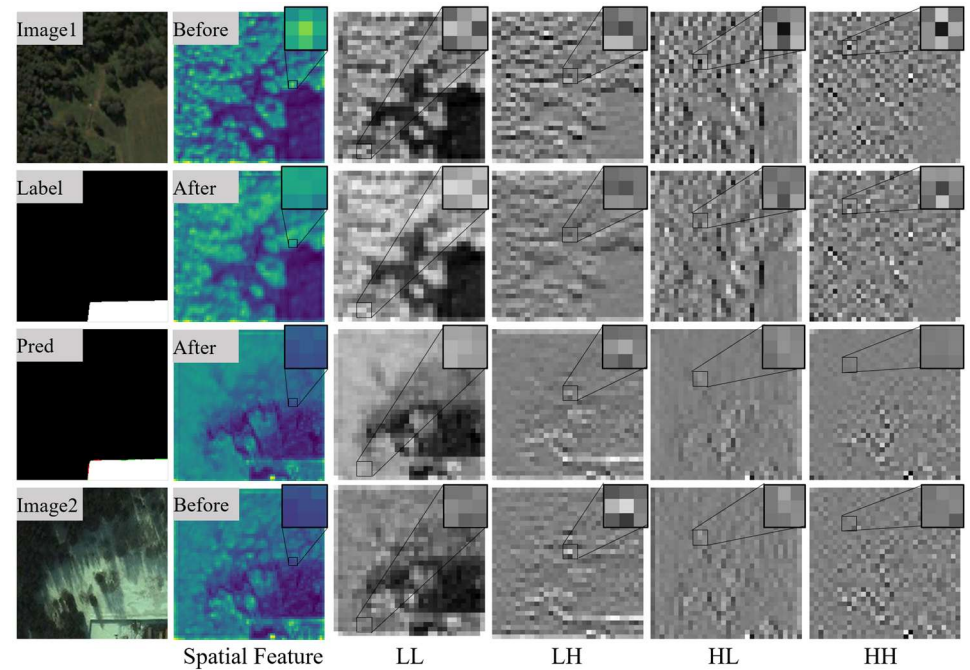
- **Suppress** pseudo-changes via frequency-domain feature interaction.
- **Enhance** real change perception using Spatio-Temporal Attention.
- **Align** deep semantics with shallow details through Gating Fusion.

# 3. Method: Frequency Branch

## Discrepancy-Aware Wavelet Interaction Module (DAWIM)



Operational workflow of the DAWIM

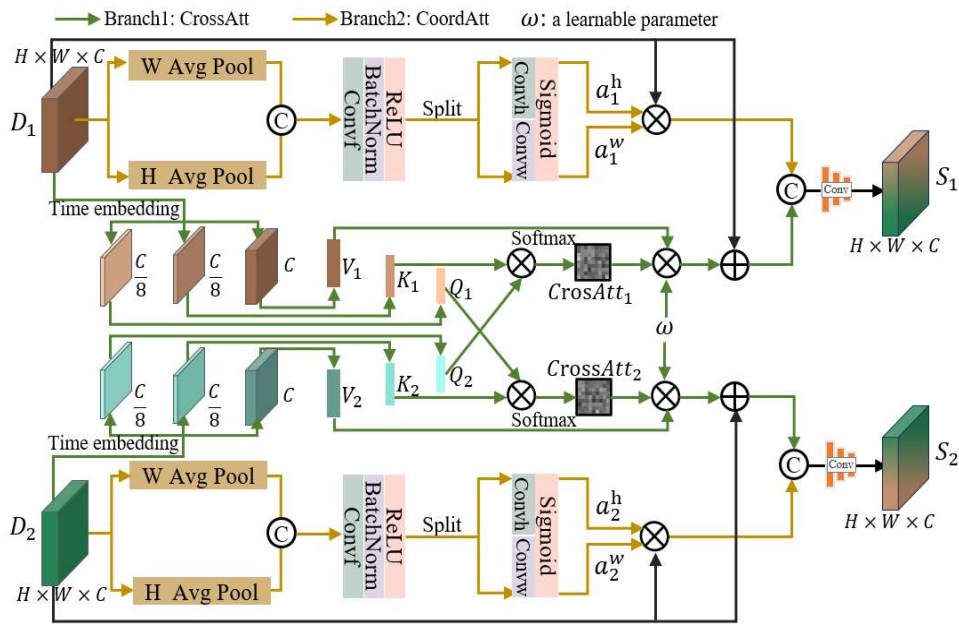


Feature modulation process of DAWIM

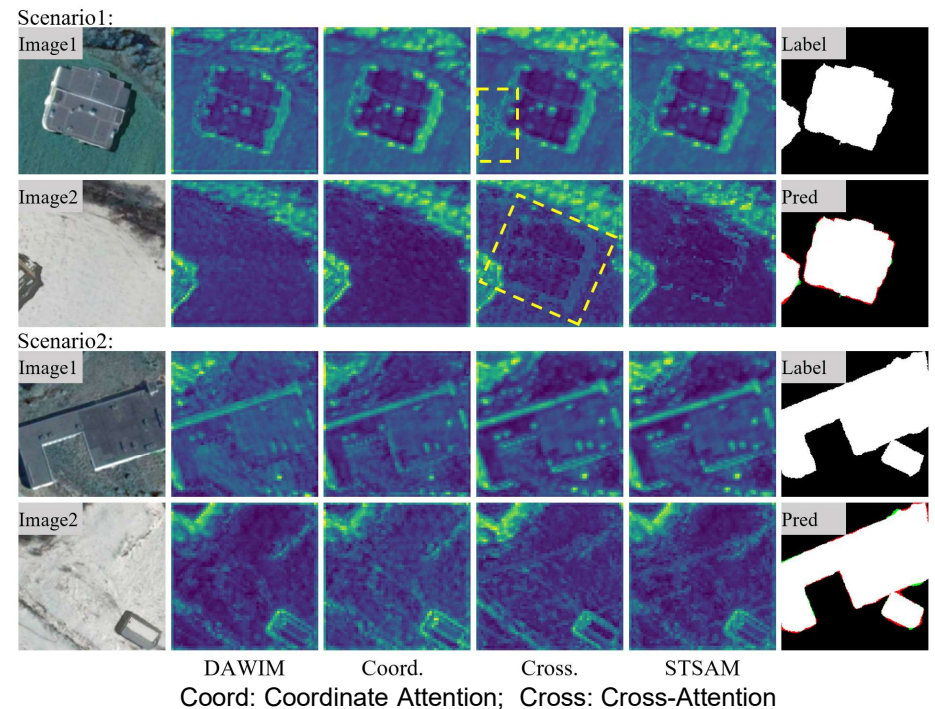
Effectively suppress pseudo-changes while preserving structural details

# 3. Method: Spatial Branch

## ■ Synergistic Temporal-Spatial Attention Module (STSAM)



Internal mechanism of the STSAM



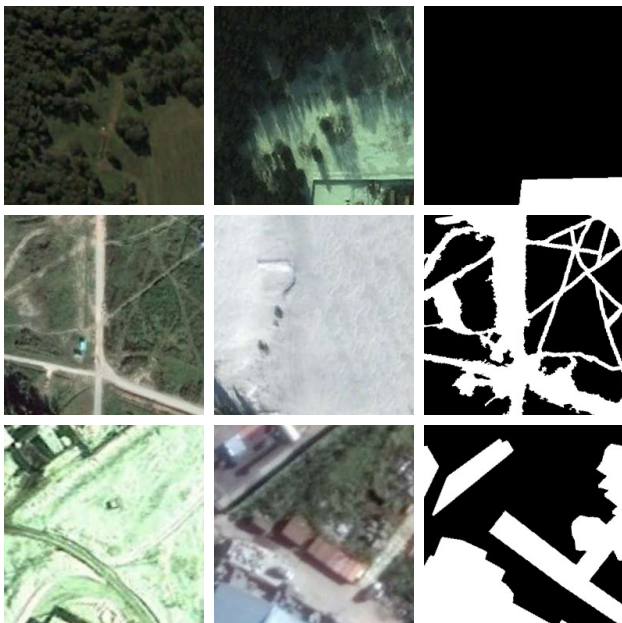
Progressive feature refinements of STSAM

Precisely enhance change region perception and detail refinement

# 4. Experiments: Datasets

## ■ Evaluated on Three Challenging Datasets

Training: 10000    Validation: 3000    Test: 3000



T1                      T2                      Label

**CDD** (Seasonal)  
16000 images  
Resolution: 0.03-1m

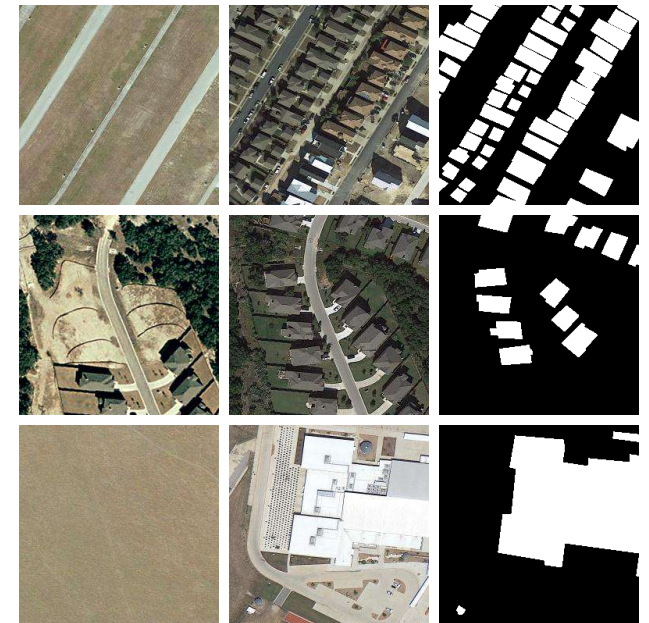
Training: 751    Validation: 214    Test: 108



T1                      T2                      Label

**GZ-CD** (Urban)  
1073 images  
Resolution: 0.5m

Training: 7120    Validation: 1024    Test: 2048



T1                      T2                      Label

**LEVIR-CD** (Building)  
10214 images  
Resolution: 0.5m

**Providing a rigorous test in handling spectral noise and spatial complexity**

# 4. Experiments: Quantitative Results

## ■ Comprehensive Performance Benchmarking

TABLE I  
COMPARISON RESULTS ON THE THREE CD DATASETS. THE BEST SCORE IS HIGHLIGHTED IN BOLD. ALL RESULTS ARE DESCRIBED AS PERCENTAGES (%)

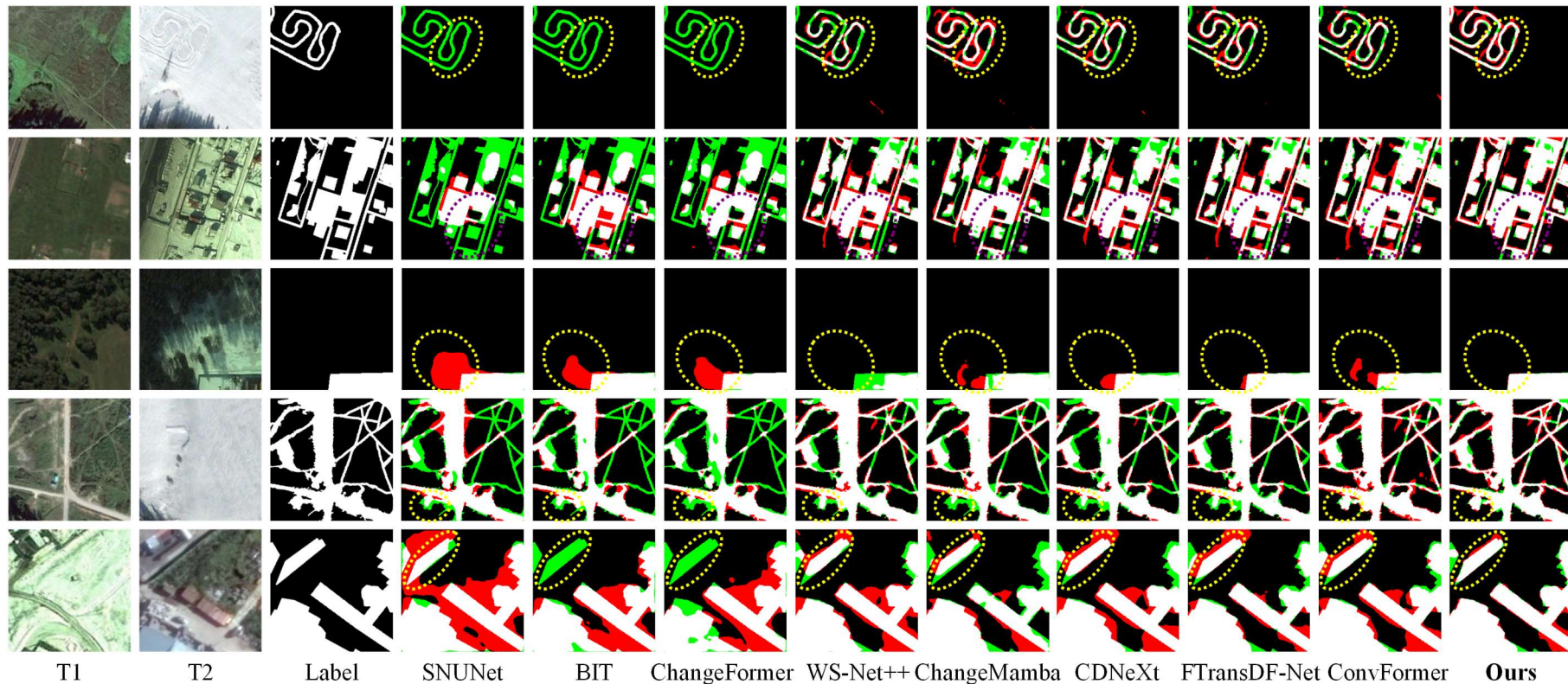
Methods	Ref.	CDD					GZ-CD					LEVIR-CD				
		Pre.	Rec.	F1	IOU	OA	Pre.	Rec.	F1	IOU	OA	Pre.	Rec.	F1	IOU	OA
SNUNet [39]	GRSL21	90.54	86.61	88.53	79.42	97.26	82.76	80.55	81.64	68.98	90.56	89.31	89.56	89.43	80.88	98.89
BIT [45]	TGRS21	90.67	86.44	88.50	79.37	97.26	82.90	80.68	81.77	69.16	90.63	89.24	89.37	89.30	80.67	98.88
ChangeFormer [43]	IGARSS22	91.54	89.31	90.41	82.50	97.68	84.59	84.28	84.43	73.06	91.90	91.97	89.03	90.48	82.62	99.02
DMINet [18]	TGRS23	92.45	90.72	91.58	84.47	97.96	89.04	86.52	87.76	78.19	93.71	92.27	88.94	90.57	82.77	99.03
AMTNet [10]	ISPRS23	92.32	92.91	92.61	86.24	98.19	88.51	85.27	86.86	76.77	93.28	91.34	89.76	90.54	82.72	99.02
WS-Net++ [20]	TGRS24	92.95	93.21	93.08	87.06	98.31	88.32	85.76	87.02	77.02	93.33	92.11	<b>90.05</b>	91.07	83.60	99.08
ChangeMamba [61]	TGRS24	92.84	93.41	93.12	87.13	98.31	89.32	87.17	88.23	78.94	93.94	91.76	89.54	90.64	82.88	99.03
CDNeXt [47]	JAG24	92.76	93.87	93.31	87.46	98.35	89.21	86.34	87.75	78.17	93.72	92.15	89.69	90.91	83.33	99.06
FTransDF-Net [26]	JAG25	93.10	94.03	93.56	87.90	98.42	89.13	87.45	88.28	79.02	93.95	92.19	89.73	90.94	83.39	99.07
ConvFormer [46]	TGRS25	<b>93.35</b>	93.19	93.28	87.41	98.36	89.11	87.58	88.33	79.11	93.97	92.14	89.69	90.90	83.32	99.06
<b>FSG-Net</b>	Ours	93.33	<b>95.01</b>	<b>94.16</b>	<b>88.96</b>	<b>98.56</b>	<b>90.85</b>	<b>88.20</b>	<b>89.51</b>	<b>81.01</b>	<b>94.61</b>	<b>92.53</b>	90.04	<b>91.27</b>	<b>83.94</b>	<b>99.10</b>

**FSG-Net achieves SOTA performance across all three datasets**

# 4. Experiments: Qualitative Results

## ■ Visualization of results on CDD Datasets

■ True Negative      □ True Positive      ■ False Negative      ■ False Positive

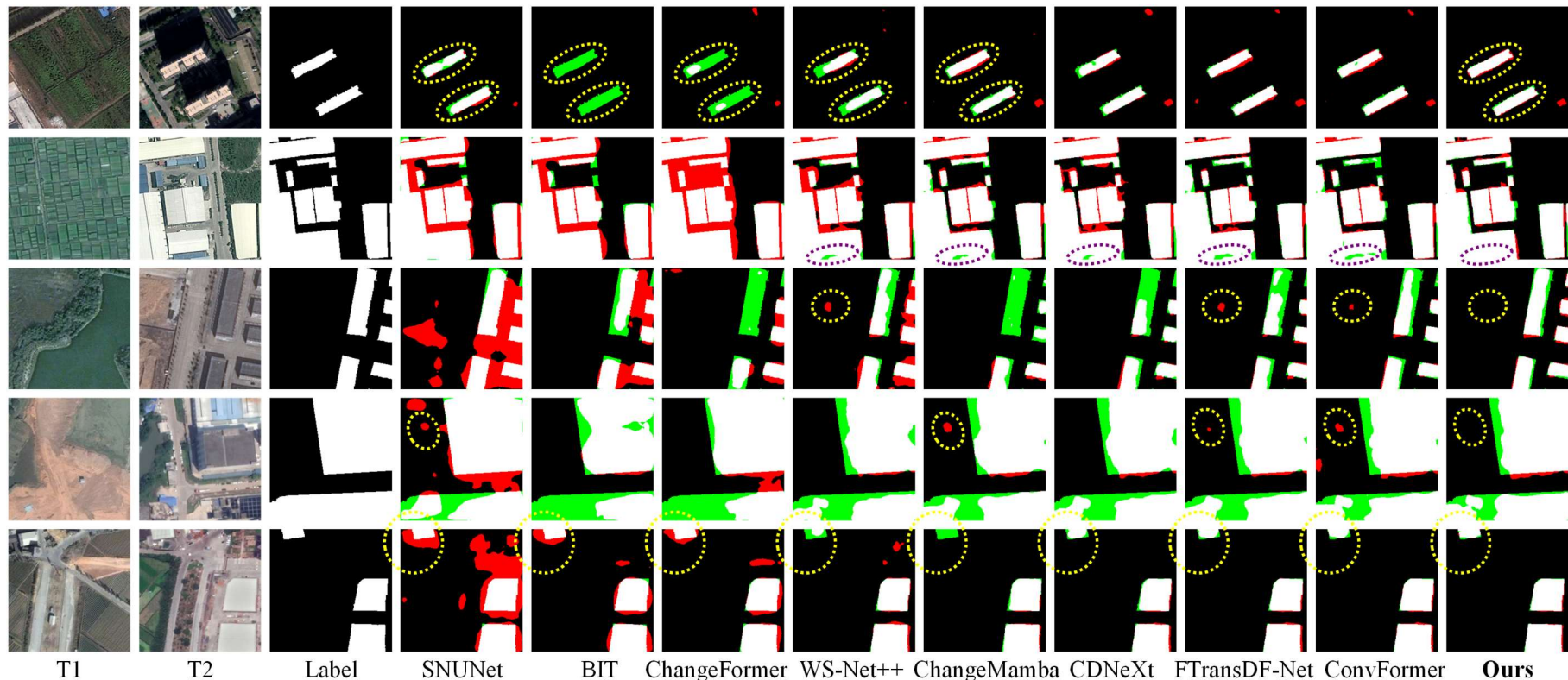


**Significant suppression of seasonal and shadow pseudo-changes**

# 4. Experiments: Qualitative Results

## ■ Visualization of results on GZ-CD Datasets

■ True Negative    □ True Positive    ■ False Negative    ■ False Positive

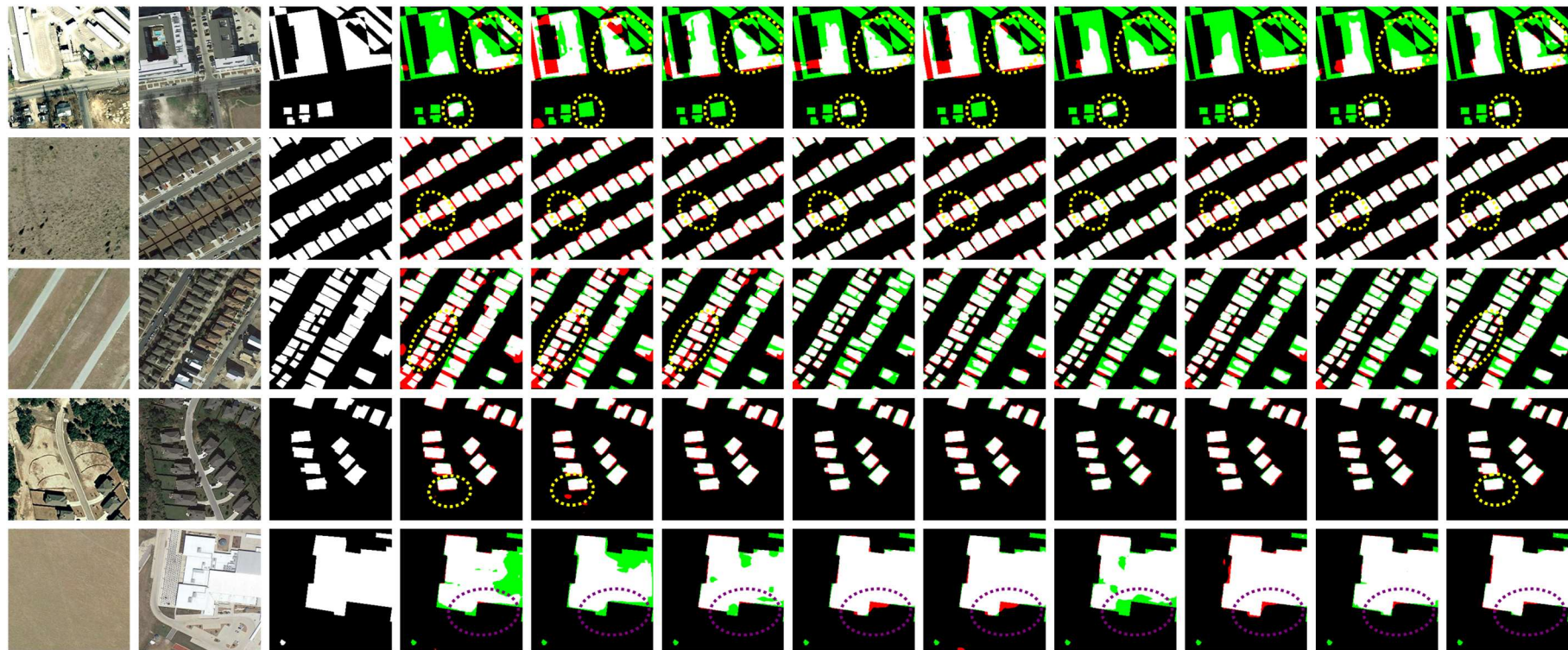


**Superior robustness to complex urban textures and multi-scale variations**

# 4. Experiments: Qualitative Results

## ■ Visualization of results on LEVIR-CD Datasets

■ True Negative      □ True Positive      ■ False Negative      ■ False Positive



**Sharp and precise building boundary delineation**

# 5. Conclusion and Outlook

## ■ What we achieved: **FSG-Net**



### A synergistic Framework

Proposed a frequency-spatial synergy pipeline to systematically disentangle semantic changes from nuisance variations.



### A “Push-Pull” Mechanism

Designed a reciprocal interaction strategy where DAWIM suppresses noise and STSAM enhances saliency.



### A New SOTA

Validated superiority across CDD, GZ-CD, and LEVIR-CD benchmarks, achieving highest F1-scores.

## ■ What is next: **Trust-CC** (will be online on arXiv soon)



### Efficiency

Focusing on lightweight network architectures and efficient feature extraction for edge device deployment.



### Multi-modality

Integrating multi-modal data (SAR, Hyperspectral) to enhance robustness in extreme imaging conditions.




### Foundation Models


Constructing knowledge graphs and rule-based constraints for semantics-aware alignment in foundation models.

**Reference:** Z. Xie et al., "FSG-Net: Frequency-Spatial Synergistic Gated Network for High-Resolution Remote Sensing Change Detection," in IEEE Transactions on Geoscience and Remote Sensing, vol. 64, Art no. 4702115, 2026, doi: 10.1109/TGRS.2026.3666124.

# Thank you for your attention!

Any Questions?

 [github.com/zxXie-Air/FSG-Net](https://github.com/zxXie-Air/FSG-Net)

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